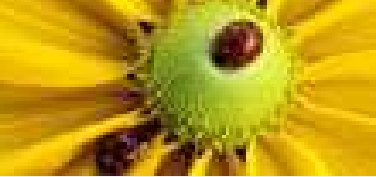


The Effect of Expected Income on Individual Migration Decisions

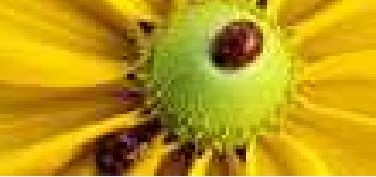
Kennan and Walker (2006)



Introduction

	Less than High School	High School	Some College	College
No. of people	1768	3534	1517	1435
Movers	423	771	376	469
Movers (%)	23.9%	21.8%	24.8%	32.7%
Moves Per Mover	2.0	1.8	1.7	1.6
Repeat moves (% of all moves)	50.6	45.9	41.3	35.7
Return Migration (% of all moves)				
Return - Home	24.0	24.1	17.5	13.4
Return - Else	12.4	7.2	5.9	3.3
Movers who return home (%)	48.7	44.5	29.8	20.9
Return-Home: % of Repeat	47.5	52.5	42.4	37.5

- Younger and more educated people are more likely to move
- Repeat moves are a large part of observed migration flows
- Return moves are a large part of observed migration flows



Introduction

- Question: What is the extent to which people move to improve their income prospects?
- Need a dynamic model that allows for multiple moves and multiple other locations.
- They build and estimate an optimal search model of migration



Model

$$V(x, \zeta) = \max_j \{v(x, j) + \zeta_j\}$$

$$v(x, j) = u(x, j) + \beta \sum_{x'} p(x'|x, j) \bar{v}(x')$$

$$\bar{v}(x) = \int V(x, \zeta) f(\zeta) d\zeta$$

ζ is drawn from a Type 1 extreme value distribution, iid across location and period and \perp to x .

$$\exp(\bar{v}(x)) = \sum_{k=1}^n \exp(v(x, k))$$

Then probability of choosing location j when the state is x is given by multinomial logit:

$$\rho(x, j) = \exp(v(x, j) - \bar{v}(x))$$



Details

If we allow agents to recall match value of every location they've visited then $J(n)^J$ states to compute (for each age and type).

So, K&W allow agents to recall only $M = 2$ most recent location matches. States = 23409.

However, 2 is a magic number.

Let $l = (l^0, l^1)$ be vector of recent locations and ω be the corresponding sequence of recent wage information. Then the state vector $x = (l, \omega, a)$



Flow payoffs

$$\tilde{u}_h(x, j) = u(x, j) + \zeta_j$$

$$u(x, j) = \alpha_0 y(l^0, \omega^0) + \sum_{k=1}^K \alpha_k Y_k(l^0) + \kappa \xi(l^0 = h) - \delta_\tau(x, j)$$

$$\delta_\tau(x, j) = (\gamma_{0,\tau} + \gamma_1 D(l^0, j) - \gamma_2 \xi(j \in A(l^0)) - \gamma_3 \xi(j = l^1) + \gamma_4 a - \gamma_5 n_j) \xi(j \neq l^0)$$



Wages

Observed wage of individual i at location j is specified as:

$$w_{ij}(a) = X_i\beta + \phi(a) + \mu_j + \nu_{ij} + \eta_i + \epsilon_{ij}(a)$$

Econometrician observes only X_i, a .

μ, ν_{ij} and η_i are observable by the agent, though ν_{ij} only for $j \in l$. Only μ_j and ν_{ij} will effect migration decisions.

They estimate μ_j , using Census data, regressing annual earnings on full set of State and age dummies.

$$L_{i\tau} = \sum_s q(i, s) \left(\prod_{t=1}^{T_i} \rho_h(x_i(t, s), j_i(t), \theta_\tau) \right)$$

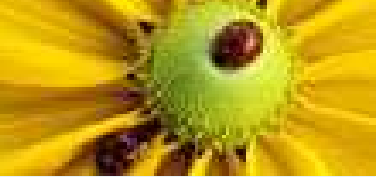


Table 2: Interstate Migration of Young White Men				
Disutility of Moving (γ_0)	7.173	3.690	4.680	4.305
	<i>0.000</i>	<i>0.543</i>	<i>0.646</i>	<i>0.645</i>
Distance (γ_1) (1000 miles, State pop centroids)		0.325	0.275	0.278
		<i>0.138</i>	<i>0.147</i>	<i>0.148</i>
Adjacent Location (γ_2)		0.643	0.665	0.669
		<i>0.153</i>	<i>0.158</i>	<i>0.159</i>
Home Premium (κ)		0.290	0.275	0.372
		<i>0.024</i>	<i>0.021</i>	<i>0.032</i>
Previous Location (γ_3)		2.820	4.380	3.767
		<i>0.235</i>	<i>0.317</i>	<i>0.327</i>
Age (γ_4)		0.094	0.112	0.095
		<i>0.021</i>	<i>0.024</i>	<i>0.024</i>
Population (γ_5) (10 million people)		0.715	0.631	0.631
		<i>0.135</i>	<i>0.132</i>	<i>0.138</i>
Stayer Probability		0.481	0	0.426
		<i>0.057</i>	---	<i>0.062</i>
Cooling (1,000 degree-days)		0.109	0.095	0.140
		<i>0.023</i>	<i>0.019</i>	<i>0.024</i>
Heating (1,000 degree-days)		0.019	0.015	0.025
		<i>0.009</i>	<i>0.008</i>	<i>0.010</i>
"Real" Income (α) (\$10,000)		-----	0.466	0.552
			<i>0.058</i>	<i>0.075</i>
Loglikelihood	-1744.88	-1309.60	-1305.44	-1287.86
Observations	5,767			
Moves	213			

Note: previous location and age are important, as are expected differences in income

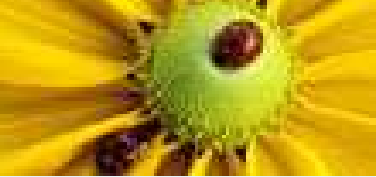


Table 3							
Moving Cost Examples							
	γ_0/α	Age	Distance	Adjacent	Population	Previous Location	Cost
Homogeneous Model							
Coefficients	\$195,361	0.1121	0.2740	0.6654	0.6304	4.3772	
Young mover		20	1	0	1	0	\$274,027
Average mover		23.5	0.715	0.338	0.759	0.300	\$229,151
Move to Previous location		20	1	0	1	1	\$91,490
Two-Type Model (mover type)							
Coefficients	\$151,637	0.0948	0.2783	0.6693	0.6315	3.7668	
Young mover		20	1	0	1	0	\$215,994
Average mover		23.5	0.715	0.338	0.759	0.300	\$176,157
Move to Previous location		20	1	0	1	1	\$58,911

Though cost is high, an agent who moves to a state with avg. wages 2σ above mean and has an match benefit of 2σ above mean would reap a lifetime benefit of \$335,000.

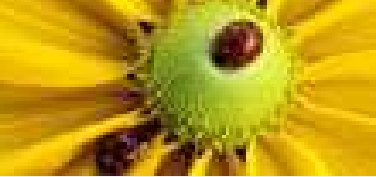


Table 5: Goodness of Fit								
Moves	Binomial		NLSY		Homogeneous Model		Two-Type Model	
	None	482.8	72.6%	544	81.80%	532083	80.01%	546251
One	154.4	23.2%	57	8.57%	69488	10.45%	49342	7.42%
More	27.80	4.2%	64	9.62%	63429	9.54%	69419	10.44%
Proportion of movers with more than one move	15.26%		52.89%		47.72%		58.45%	
Total observations	665		665		665000		665012	

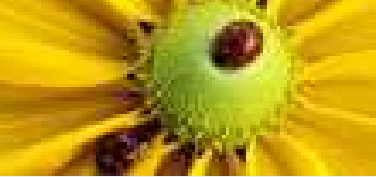
Binomial probability is 3.69% - picked to match number of moves per person-year in data.

Table 6: Return Migration Statistics			
	NLSY	Homogeneous Model	Two-Type Model
Proportion of Movers who			
Return home	33.8%	32.7%	33.7%
Return elsewhere	5.6%	7.1%	7.4%
Move on	60.6%	60.1%	58.8%
Proportion who <i>ever</i>			
Leave Home	15.3%	15.4%	14.8%
Move from not-home	41.7%	58.2%	43.3%
Return from not-home	23.6%	31.7%	28.6%

Human capital and age

	All		Not At Home ^a		At Home	
Age	N	Migration Rate	N	Migration Rate	N	Migration Rate
20	677	4.73%	74	21.62%	603	2.65%
21	637	4.87%	74	14.86%	563	3.55%
22	609	5.09%	81	19.75%	528	2.84%
23	569	3.51%	83	13.25%	486	1.85%
24	587	4.09%	83	15.66%	504	2.18%
25	533	4.69%	79	12.66%	454	3.30%
26	512	4.49%	80	17.50%	432	2.08%
27	465	1.94%	73	9.59%	392	0.51%
28	381	1.57%	57	5.26%	324	0.93%
29	307	1.63%	51	3.92%	256	1.17%
30	242	1.65%	38	7.89%	204	0.49%
31	149	2.01%	21	9.52%	128	0.78%
32	81	0.00%	12	0.00%	69	0.00%
33	18	0.00%	1	0.00%	17	0.00%
All	5,767	3.69%	807	13.38%	4,960	2.12%

^aAt Home means living now in the State of residence at age 14.



Note: Large drop in migration rates over life-cycle. 3 reasons for this in model:

1. Workers are poorly sorted initially, so workers may want to move initially.
2. Decreasing benefit over life-cycle of incurring fixed cost of moving (similar to education/humancapital models).
3. Age related moving costs.

The fact that (3) is significant indicates that human capital effect cannot completely explain migration.