

# LOVE THY NEIGHBOR?

An empirical test of neighborhood ethnicity change and Schelling behavior

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- **Motivation:** Residential segregation in Denmark and Schelling's model
- **Research Question:** Does the ethnicity of your nearest neighbor affect propensity to move?
- **Methods:** Nearest-neighbor research design with comprehensive administrative data
- **Results:** Asymmetry in residential responses based on ethnicity
- **Heterogeneity Analysis:** SES differences in Schelling behavior
- **Conclusion:** Evidence of individually motivated segregation

*Definitions:*

1. Native households: All members are of Danish origin
2. Non-Western households: At least 1 member is of non-Western origin
3. Western households: At least 1 member is of Western origin (but no non-Western household members)

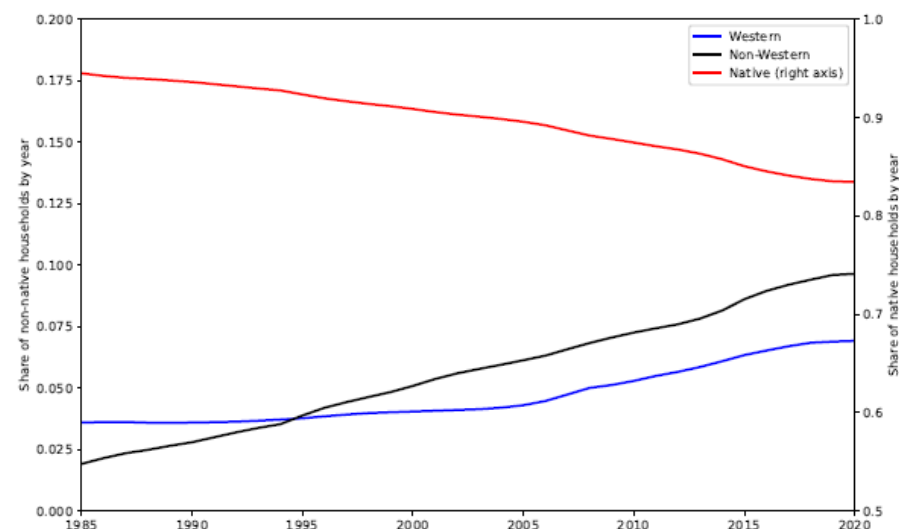
# INTRODUCTION

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- What should go in appendix and what should be in prez?
- Exclude data sources(?)
- Spend time on spatial patterns. This
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- Denmark has transformed from a relatively homogeneous society to increasing ethnic diversity
- Non-Western households grew from ~2% in 1985 to ~10% by 2020
- Limited empirical evidence on how ethnic background directly influences residential sorting

Figure 1: Distribution of households (1985-2020)



Note: This is based on own calculations and shows the distribution of households by type as defined in section 1.1. I sample households that are present at December 31st in each year from 1985.

- **(Schelling, 1971)** proposed that neighborhoods may “tip” when minority share reaches a threshold
- Even with relatively tolerant preferences toward diversity
- Three types of segregation:
  1. Organized segregation (e.g., historical Jim Crow laws)
  2. Economically induced segregation (clustering by income/education)
  3. **Individually motivated segregation** ← **Focus of this paper**
- Schelling’s key insight: Small individual preferences can lead to macro-level segregation

# METHODS

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Modeling a household's decision to stay or move in a neighborhood that evolves over time:

$$U_{i,j,t} = f(Z_{i,t}, X_{j,t}, \xi_{j,t}) + \sum_k g(Z_i, Z_{k,t}, D_{i,k}) + \varepsilon_{i,j,t}$$

Where:

- $f(\cdot)$ : Utility from neighborhood amenities
- $g(\cdot)$ : Utility from characteristics of each neighbor  $k$  at distance  $D_{i,k}$
- $Z_i$ : Observable household attributes
- $X_j$ : Observable neighborhood attributes
- $\xi_j$ : Unobservable neighborhood attributes
- $\varepsilon_{i,j,t}$ : Idiosyncratic preferences



$$V_{i,j,t} = f(Z_{i,t}, X_{j,t}, \xi_{j,t}, \alpha) + \sum_k g(Z_{i,t}, Z_{k,t}, D_{i,k}, \beta) + \delta E[V_{i,j,t+1}] + \varepsilon_{i,j,t}$$

## Key identification challenges:

- Unobserved neighborhood amenities
- Dynamic preferences (expectations of future changes)
- Selection effects (who moves where is not random)

## Innovative approach from (Bayer *et al.*, 2022):

Compare households within the same neighborhood who receive different-type neighbors:

- **Treatment group:** Households with new different-type neighbors among their 3 nearest neighbors
- **Control group:** Households with new different-type neighbors “just down the road” (ranks 4-6)

$$Y_{i,j,t} = \beta_1 I[e', k = n_{nearest}] + \beta_2 I[e', k = n_{near}] + \beta_3 I[e', k = n_{close}] + \gamma Z_{i,j,t} + \omega_{j,t} + \varepsilon_{i,j,t}$$

Parameter of interest:

$$\beta_1 - \beta_2$$

This design addresses key identification challenges by comparing households experiencing same neighborhood conditions but different micro-geography of new neighbors.

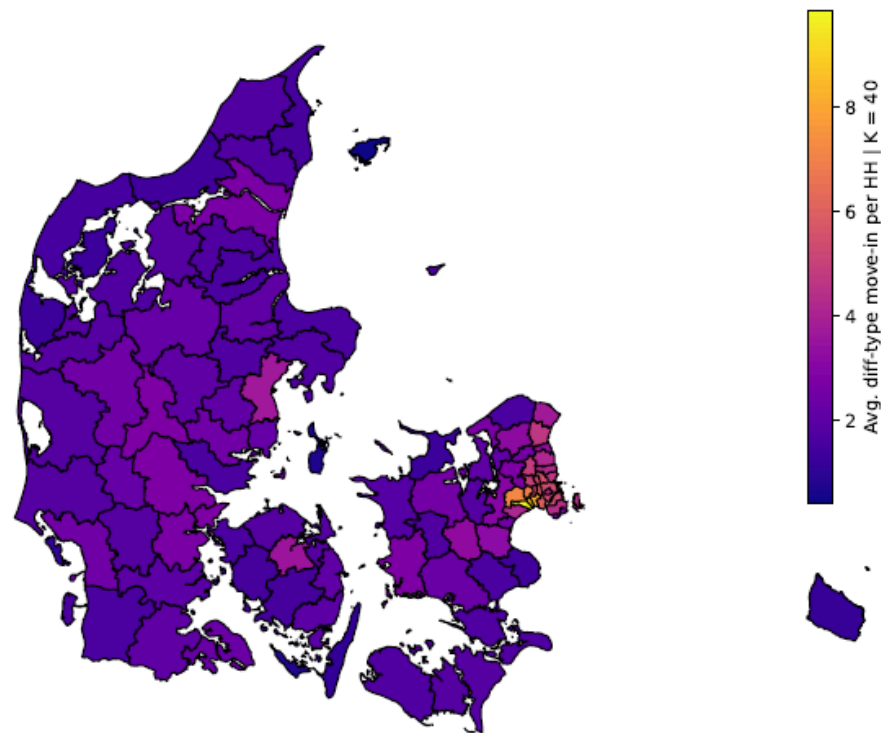
# DATA

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## Key spatial patterns:

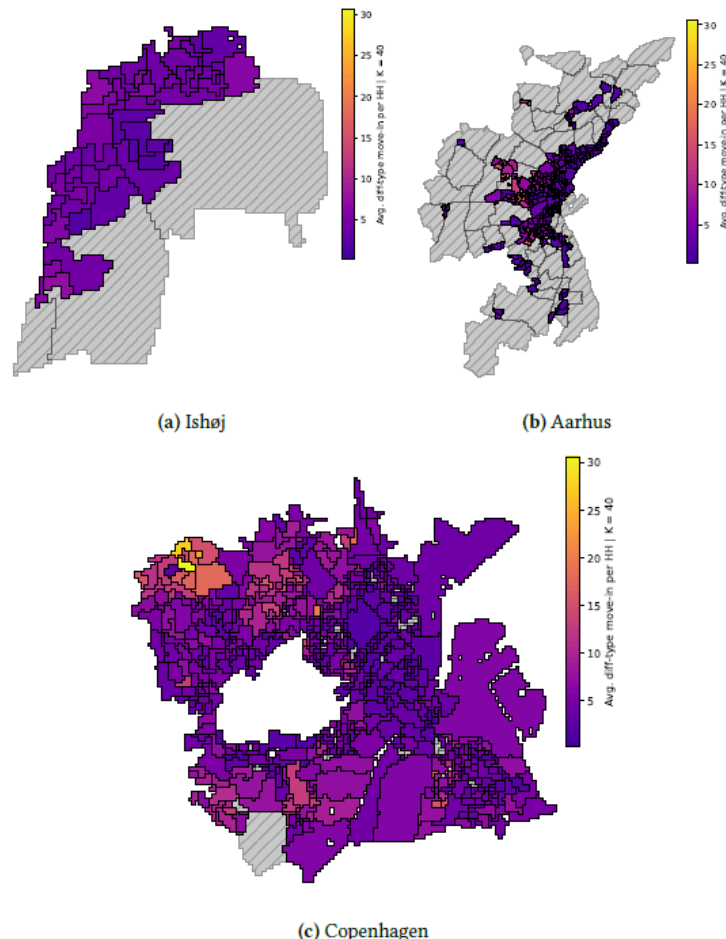
- Clear east-west and urban-rural divide
- Concentration in Copenhagen and surroundings
- Highest incidence in Ishøj (~9 new different-type neighbors)
- Copenhagen (~6), Aarhus and Odense (~4)

Figure 3: Incidence of new different-type neighbors (1985-2020)



*Note:* The figure show the variation in receiving a new-non Western neighbors within the 40 closest parcels for native households. Municipal borders correspond to the ones imposed by "Kommunalreformen" in 2007. Household types are split up in to three types, see section 1.1 for more details.

Figure 4: Incidence of new different-type neighbors at the neighborhood level (1985-2020)



## Within-city variation:

- Some Copenhagen neighborhoods: 30+ new non-Western neighbors
- Other Copenhagen neighborhoods: <2 new non-Western neighbors

## 3 key observations from summary statistics:

1. Treated households show higher mobility: 23-24% vs. 19-20% for “control” households
2. Treated native households have lower wealth (48,500 DKK vs. 81,000 DKK) and income
3. Non-Western households are better educated on average (by ~2 years)

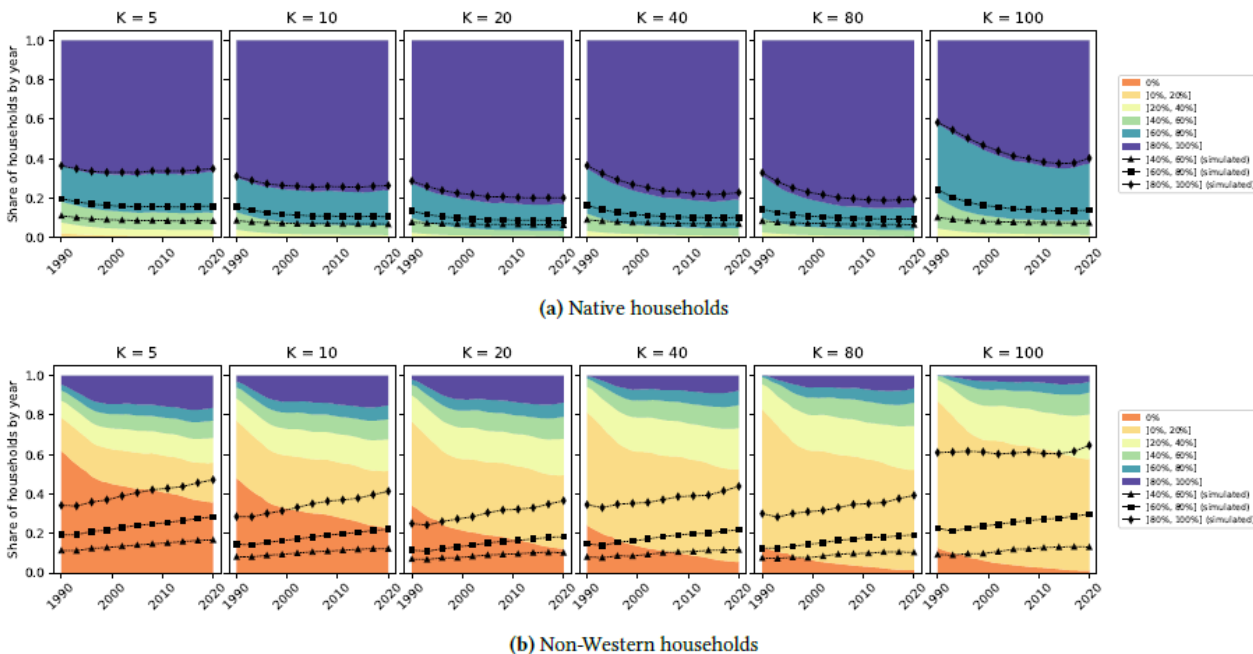
Table 1: Summary statistics

	Native households						Non-Western households					
	All		Nearest		Close		All		Nearest		Close	
Household characteristics												
Move within 2 years	17.45	(37.95)	23.19	(42.21)	19.74	(39.80)	19.14	(39.34)	24.01	(42.71)	19.34	(39.50)
Real inc. (1000s) DKK	344.11	(120.22)	337.16	(119.58)	341.07	(120.29)	313.13	(111.14)	317.37	(114.96)	314.62	(112.27)
Real net wealth (1000s) DKK	81.03	(202.68)	48.51	(185.87)	62.08	(193.44)	45.52	(163.75)	41.34	(160.52)	46.95	(165.29)
Employed	0.86	(0.34)	0.83	(0.37)	0.84	(0.37)	0.83	(0.38)	0.84	(0.37)	0.83	(0.37)
Years of education	10.15	(6.24)	10.03	(6.26)	10.13	(6.25)	12.12	(5.34)	12.13	(5.47)	12.10	(5.37)
Distance to neighbor			3.20	(4.29)	33.15	(20.69)			2.81	(3.70)	30.60	(18.15)
Household size	1.81	(1.06)	1.63	(0.94)	1.70	(0.99)	2.14	(1.29)	1.96	(1.20)	2.13	(1.27)
Oldest household member	44.25	(8.93)	43.51	(9.03)	44.18	(9.00)	43.37	(8.71)	41.84	(8.61)	43.26	(8.69)
Neighborhood characteristics												
Population density	7418.23	(7644.95)	8969.51	(8151.09)	8387.60	(7945.02)	9198.14	(8155.82)	9667.16	(8521.05)	9331.84	(8326.29)
Native share	0.89	(0.10)	0.81	(0.14)	0.83	(0.13)	0.76	(0.18)	0.78	(0.16)	0.78	(0.16)
Non-Western share	0.08	(0.09)	0.15	(0.14)	0.13	(0.12)	0.19	(0.17)	0.17	(0.15)	0.18	(0.16)
Real income (median)	322.85	(32.94)	318.41	(35.49)	320.60	(34.61)	317.40	(35.96)	319.73	(36.69)	318.51	(35.72)
Real net wealth (median)	42.74	(55.89)	27.58	(44.34)	32.01	(48.11)	27.87	(44.55)	27.54	(42.75)	27.90	(43.95)
N	33,496,551		761,842		4,671,062		3,330,242		414,103		1,443,373	

Note: This table shows presents summary statistics for households “at-risk” of receiving a different-type neighbor. Standard deviations in parenthesis. Income and wealth are equalised to facilitate comparison between households of different size and composition. The All column denotes quarter-by-year observation for the sample of household defined in section 3.6. The Nearest (“treated”) and Close (“control”) columns denote instances, where a household experienced a new different-type among their  $K \in [1, 2, 3]$  nearest neighbors or close neighbors ( $K \in [4, 5, \dots, 40]$ ).

- Given demographic context, how has residential sorting developed over time?
- Natives:**
  - 1990 [K=100 nearest]: 40 percent had between 80-100 same-type neighbors.
  - 2020 [K=100 nearest]: 60 percent had between 80-100 same-type neighbors.
- Quick and dirty* counterfactual simulation with fixed 1990-distribution show little-to-no change in sorting patterns.
- Schelling behavior?**

Figure 5: Same type neighbor by  $K$ -nearest proximity (1990-2020)



# RESULTS

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Table 4: Estimates of Schelling behavior (native households)

	Move within 2 years (=100)					
	(1)	(2)	(3)	(4)	(5)	(6)
New diff neighbor $k_{nearest} \text{ v } k_{near}$	0.357*** (0.084)	0.368*** (0.084)	0.354*** (0.081)	0.328*** (0.081)	0.309*** (0.081)	0.320*** (0.081)
N	5,365,811	5,365,811	5,365,811	5,365,811	5,365,811	5,365,811
Neighborhood-by-quarter FE	X	X	X	X	X	X
Mean of dependent variable	20.23	20.23	20.23	20.23	20.23	20.23
Number of neighborhoods	3444	3444	3444	3444	3444	3444
Income		X	X	X		X
Wealth					X	X
Tenure			X	X	X	X
Age				X	X	X

\* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001. Standard errors (in parenthesis) are clustered at the neighborhood level. The table reports the estimate of  $\beta_1 - \beta_2$  from equation 6. Table A.3 contains the complete set of coefficients.

Table 5: Estimates of Schelling behavior (non-Western households)

	Move within 2 years (=100)					
	(1)	(2)	(3)	(4)	(5)	(6)
New diff neighbor $k_{nearest} \text{ v } k_{near}$	0.063 (0.129)	0.062 (0.129)	0.155 (0.127)	0.097 (0.126)	0.097 (0.126)	0.096 (0.126)
N	1,795,109	1,795,109	1,795,109	1,795,109	1,795,109	1,795,109
Neighborhood-by-quarter FE	X	X	X	X	X	X
Mean of dependent variable	19.96	19.96	19.96	19.96	19.96	19.96
Number of neighborhoods	3332	3332	3332	3332	3332	3332
Income		X	X	X		X
Wealth					X	X
Tenure			X	X	X	X
Age				X	X	X

\* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001. Standard errors (in parenthesis) are clustered at the neighborhood level. The table reports the estimate of  $\beta_1 - \beta_2$  from equation 6. Table A.4 contains the complete set of coefficients.

- Increase moving propensity by ~0.3 percentage points when receiving a new non-Western neighbor
- 1.6% increase relative to baseline exit rate
- Effect stable across specifications
- Robust to controls for income, wealth, age, tenure

- Show substantially smaller response: 0.06-0.1 percentage points
- ~0.5% relative to baseline exit rate
- Not statistically significant
- Suggests they are unaffected by identity of new native neighbors

## SES definitions:

- **Low SES:** Income  $< 200,000$  DKK, outside labor market or  $\leq 11$  years of education
- **High SES:** Income  $\geq 600,000$  DKK, employed full-time or  $\geq 18$  years of education

## Key findings:

- Schelling behavior primarily driven by low-SES native households responding to low-SES non-Western households
- Effect size:  $\sim 0.56$  percentage points or  $\sim 2.8\%$  increase from baseline exit rate
- Nearly twice the magnitude observed in full sample
- Very rare for low-SES native households to receive high-SES non-Western neighbors and vice versa
- Confirms powerful residential sorting at neighborhood level

**Danish findings vs. (Bayer *et al.*, 2022) U.S. results:**

Context	Magnitude
Denmark (Native)	1.6% above baseline
Denmark (Non-Western)	0.5% above baseline
U.S. (White)	4% above baseline
U.S. (Black)	6% above baseline

**Possible explanations for differences:**

- Institutional variation in housing market and integration policies
- Different neighborhood contexts (urban/dense vs. suburban)
- Stronger “type 1 & 2” (institutional & economical) segregation?

# CONCLUSION

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1. Native Danish households increase moving propensity by 1.6% when receiving non-Western neighbors
2. Non-Western households show no significant response to new native neighbors
3. Heterogeneity by SES: Low-SES native households responding to low-SES non-Western neighbors show strongest effects (2.8%)
4. Spatial decay of effects: Moving response decreases monotonically with distance to new different-type neighbors
5. Magnitude in Denmark (1.6%) more modest than in U.S. context (4-6%)

## **Contributions to segregation research:**

- Causal evidence of individually motivated segregation as theorized by Schelling (1971)
- Demonstration of asymmetric responses in the European welfare state context
- Socioeconomic gradient in responses highlighting intersection of ethnicity and economic resources
- Evidence that Schelling mechanisms operate across different settings, but with context-specific magnitude and symmetry

## **Policy implications:**

- Integration efforts may need to account for micro-geography of neighborhood mixing

- Targeted interventions may be more effective for low-SES populations
- Understanding asymmetric responses could inform more effective housing policies

Thank you for your  
attention!

Questions?



## REFERENCES

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Bayer, P. *et al.* (2022) “Distinguishing Causes of Neighborhood Racial Change: A Nearest Neighbor Design,” *Social Science Research Network* [Preprint]. Available at: <https://doi.org/10.3386/w30487>.

Schelling, T.C. (1971) “Dynamic models of segregation,” *Journal of mathematical sociology*, 1(2), pp. 143–186.

## APPENDIX

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$$X_{i,j,t} = \varphi_1 I[e', k = n_{nearest}] + \varphi_2 I[e', k = n_{near}] + \varphi_3 I[e', k = n_{close}] + \omega_{j,t} + \varepsilon_{i,j,t}$$

**Table 2:** Balance test (native)

	Income (1,000)	Net wealth (1,000)	Oldest HH member (years)	Tenure (days)	Employed	Educ. length (years)	HH size
New diff neighbor $k_{nearest} \text{ v } k_{near}$	-1.408*** (0.254)	-1.903*** (0.402)	-0.043* (0.020)	1.936 (5.999)	-0.002** (0.001)	-0.035* (0.014)	-0.009*** (0.002)
N	5,365,811	5,365,811	5,365,811	5,365,811	5,365,811	5,365,811	5,365,811
Neighborhood-by-quarter FE	X	X	X	X	X	X	X
Mean of dependent variable	341.26	60.61	44.05	2646.87	0.84	10.11	1.70
Number of neighborhoods	3444	3444	3444	3444	3444	3444	3444

\* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001 Standard errors (in parenthesis) are clustered at the neighborhood level. The table reports the estimate of  $\phi_1 - \phi_2$  from equation 11. Table A.1 contains the complete set of coefficients.

**Table 3:** Balance test (non-Western)

	Income (1,000)	Net wealth (1,000)	Oldest HH member (years)	Tenure (days)	Employed	Educ. length (years)	HH size
New diff neighbor $k_{nearest} \text{ v } k_{near}$	-0.065 (0.347)	0.171 (0.522)	-0.061* (0.028)	31.109*** (7.656)	0.001 (0.001)	-0.011 (0.018)	0.001 (0.004)
N	1,795,109	1,795,109	1,795,109	1,795,109	1,795,109	1,795,109	1,795,109
Neighborhood-by-quarter FE	X	X	X	X	X	X	X
Mean of dependent variable	315.64	46.50	42.94	2298.84	0.84	12.11	2.11
Number of neighborhoods	3332	3332	3332	3332	3332	3332	3332

\* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001 Standard errors (in parenthesis) are clustered at the neighborhood level. The table reports the estimate of  $\phi_1 - \phi_2$  from equation 11. Table A.1 contains the complete set of coefficients.

## **Alternative specifications:**

- Combining all control distances into a single category
- Varying distance thresholds for nearest neighbors
- Different neighborhood definitions

## **Results remain consistent across specifications:**

- Spatial decay of effects provides additional support for Schelling mechanism
- Moving response decreases monotonically with distance to new different-type neighbors
- Effects primarily concentrated within 25 meters

## Simple agent-based model:

- Agents of two types randomly allocated on grid
- Agents move if share of different-type neighbors exceeds tolerance threshold
- Even with modest tolerance thresholds, segregation emerges
- This visualization demonstrates how small individual preferences

can lead to significant macro-level segregation patterns

