

Models of migration

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Introduction

- Migration : change of residence (relocation)
- Migration is situated in time and space
 - Conceptual issues
 - Space: administrative boundaries
 - Time: duration of residence or intention to stay
- Measurement of migration
 - Event: ‘migration’
 - Person: ‘migrant’

Data types applied to life-history data: the case of migration

Data types applied to migration

- Micro-data: data on individuals or households
 - **Status data:**
 - Current status (*state occupied*)
 - **migrant status** (e.g. ever migrated / never migrated in given period)
 - current place (region) of residence
 - Place of residence at two points in time: **transition data** (*migrant data*)
 - Time interval of fixed length: e.g. census and 5 years prior
⇒ “Where did you live 5 years ago?”
 - Time interval variable: e.g. census and place of birth
⇒ “Place of birth”
 - Place of residence at 3 or more points in time

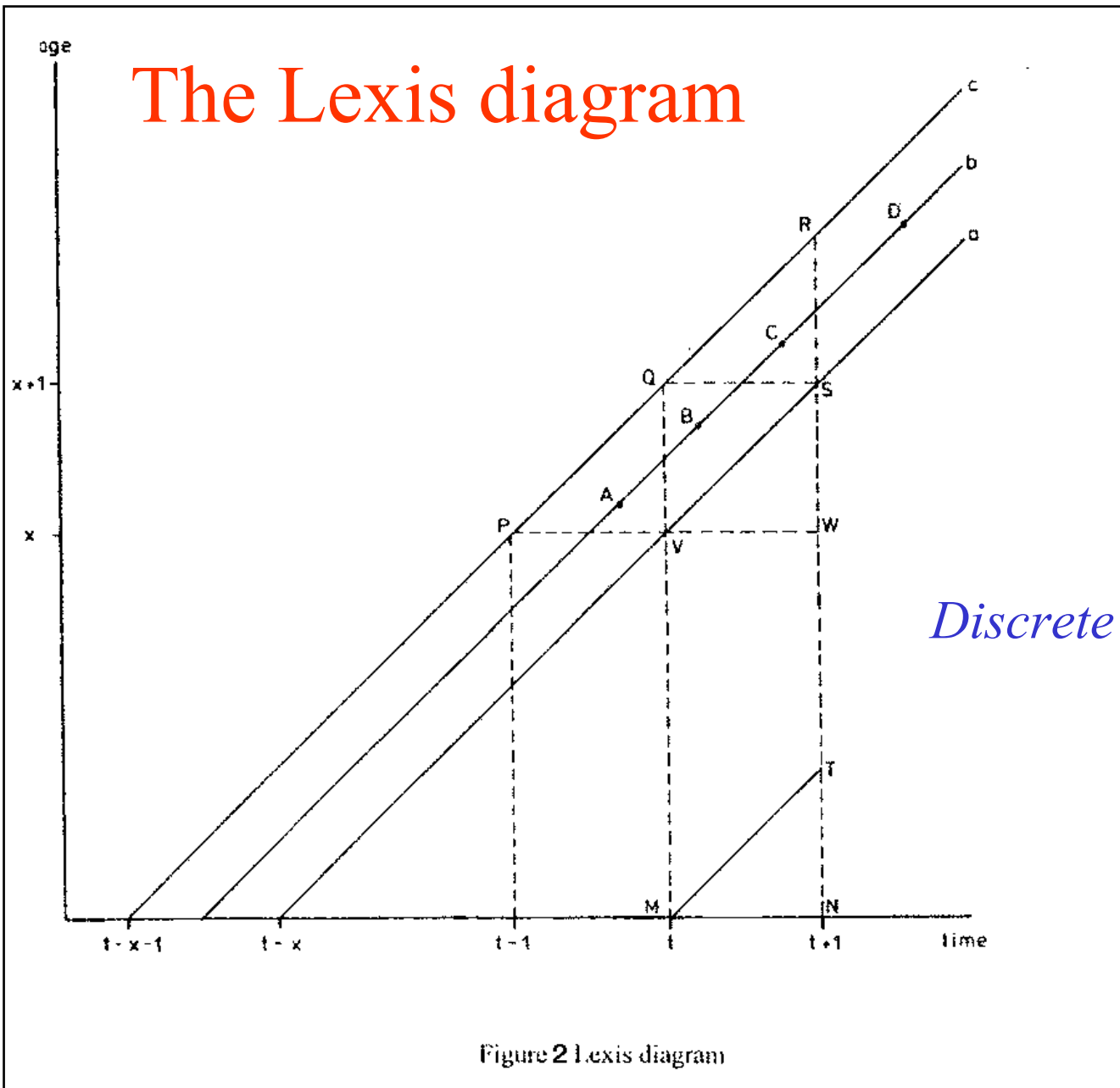
Data types applied to migration

- Micro-data:
 - **Event data** : *migration data (movement data)*
 - Migration during given period (yes/no): migrant status
 - Ever migrated?
 - Number of migrations (**quantum**)
 - Timing of migration (**tempo**)
 - Time scale: calendar time, age, process time (time since event-origin)
 - Measurement of time: exact time, time interval (discrete time, e.g. month, year)
 - Timing of all migrations vs timing of last migration

Data types applied to migration

- Grouped data: data on groups of individuals or households (actors)
 - **Status data:**
 - Current status: number of actors (subjects) in given status
 - Number of actors by place of residence at two points in time: **transition data (migrant data)** *CENSUS*
 - Number of actors by place of residence at 3 or more points in time
 - **Event data:**
 - Number of events during given period *POP. REGISTER*

The Lexis diagram



Discrete age/time

Figure 2 Lexis diagram

Individual as 'carrier' of attributes

Life course as a sequence of attributes

- Personal attribute at age x or time t
- Attributes change in the course of life: events
- Describe life course
 - By attributes at each age: **status-based approach**
 - By initial attributes and changes in attributes: **event-based approach**
- Changes (and events) occur in *continuous time* but they are often measured in *discrete time*

State-space approach to life histories

- Attribute = state
- Set of possible attributes = state space
- Attribute at age x = state occupied at age x
 - State occupancy
- Change in attribute = state transition
 - Direct transition
 - Discrete-time transition
 - Attributes of transitions: state of origin, state of destination, reason
- Timing of state transitions
 - *Age structure of migration* (age/duration dependence)
- Dependence of destination on origin
 - *Spatial structure of migration*

Probability models of migration

- Risk indicators: risk of a transition
 - Number of transitions during unit interval: **counts**
 - How likely is a transition in unit interval: **probability**
 - Timing of transition: transition **rate**
- Underlying random mechanism
 - Count data: Poisson models
 - Probability (or proportion): logit model and logistic regression
 - Rate: transition rate model
 - $\text{Rate} = \text{occurrences} / \text{exposure}$

Model 1: state occupancy

- $Y_k(\mathbf{x})$ State occupied at \mathbf{x}
- ${}_k\pi_i(\mathbf{x}) = \Pr\{Y_k(\mathbf{x})=i\}$ State probability
 - Identical individuals: ${}_k\pi_i(\mathbf{x}) = \pi_i(\mathbf{x})$ for all k
 - Individuals differ in some attributes:
 - ${}_k\pi_i(\mathbf{x}) = \pi_i(\mathbf{x}, Z)$, $Z =$ covariates
 - Prob. of residing in i region by region of birth
- Statistical inference: MLE of $\pi_i(\mathbf{x})$
 - Multinomial distribution

$$\Pr\{N_1 = n_1, N_2 = n_2, \dots\} = \frac{m!}{\prod_{i=1}^I n_i!} \prod_{i=1}^I \pi_i^{n_i}$$

Model 1: state occupancy

- Statistical inference: MLE of state probability π_i

- Multinomial distribution

$$\Pr\{N_1 = n_1, N_2 = n_2, \dots\} = \frac{m!}{\prod_{i=1}^I n_i!} \prod_{i=1}^I \pi_i^{n_i}$$

- Likelihood function $L = \prod_{i=1}^I \pi_i^{n_i}$

- Log-likelihood function $l = \ln(L) = \sum_{i=1}^I n_i \ln(\pi_i)$

- MLE $\hat{\pi}_i = \frac{n_i}{m}$

- Expected number of individuals in i : $E[N_i] = \pi_i m$

Model 1: State occupancy with covariates

$$\log \text{it}[\pi_i(Z)] = \ln \frac{\pi_i(Z)}{1 - \pi_i(Z)} = \eta_i = \beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \beta_{i3}Z_3 + \dots$$

$$\pi_i = \frac{\exp(\eta_i)}{\exp(\eta_1) + \exp(\eta_2) + \dots + 1 + \dots} = \frac{\exp(\eta_i)}{\sum_{j=1}^I \exp(\eta_j)}$$

multinomial logistic regression model

Model 2: Transition probabilities

- State probability $\pi_i(\mathbf{x}, \mathbf{Z}) = \Pr\{Y_k(\mathbf{x}, \mathbf{Z})=i \mid \mathbf{Z}\}$

- Transition probability

$$\Pr\{Y(x+1) = j \mid Y(x), Y(x-1), \dots; \mathbf{Z}\} = \Pr\{Y(x+1) = j \mid Y(x); \mathbf{Z}\}$$

$$\Pr\{Y(x+1) = j \mid Y(x) = i\} = p_{ij}(x)$$

discrete-time transition probability

Migrant data; Option 2

- Transition probability as a logit model

$$\log \text{it}[\pi_j(x+1)] = \beta_{j0} + \beta_{j1} Y_i(x) \quad p_{ij}(x) = \frac{\exp[\beta_{j0} + \beta_{j1} Y_i(x)]}{\sum_{r=1}^I \exp[\beta_{j0} + \beta_{j1} Y_r(x)]}$$

Model 2: Transition probabilities

- Transition probability as a logit model

$$\log \text{it}[\pi_j(x+1)] = \beta_{j_0}(x) + \beta_{j_1}(x)Y_i(x)$$

$$p_{ij}(x) = \frac{\exp[\beta_{j_0}(x) + \beta_{j_1}(x)Y_i(x)]}{\sum_{r=1}^I \exp[\beta_{j_0}(x) + \beta_{j_1}(x)Y_r(x)]}$$

with $\beta_{j_0}(x) = \text{logit of residing in } j \text{ at } x+1 \text{ for reference category}$
(not residing in i at x) and $\beta_{j_0}(x) + \beta_{j_1}(x) = \text{logit of residing in}$
 $j \text{ at } x+1 \text{ for resident of } i \text{ at } x$.

Model 2: Transition probabilities with covariates

Illustration 1 - Micro-data

- Covariate: region of birth

$$p_{ij}(x) = \frac{\exp[\eta_{ij}(x)]}{\sum_{r=1}^I \exp[\eta_{ij}(x)]}$$

with $\eta_{ij}(x) = \beta_{ij0}(x) + \beta_{ij1}(x)Z_1 + \beta_{ij2}(x)Z_2 + \beta_{ij3}(x)Z_3 + \dots$

e.g. $Z_k = 1$ if k is region of birth ($k \neq i$); 0 otherwise.

$\beta_{ij0}(x)$ is logit of residing in j at $x+1$ for someone who resides in i at x and was born in i .

multinomial logistic regression model

Model 2: Transition probabilities with covariates

Illustration 1 - Macro-data

- Covariate: underfive (or infant)
migration probability

$$p_{ij}(x) = \frac{\exp[\eta_{ij}(x)]}{\sum_{r=1}^I \exp[\eta_{ij}(x)]}$$

with

$$\eta_{ij}(x) = \beta_0(x) + \beta_1(x) p_{ij}(-5)$$

Rogers, Muhidin, Jordan, Lea (2004, p. 8):

linear model with regression coefficients
independent of x

$$p_{ij}(x) = a + b p_{ij}(-5)$$

Transition rates

$$\mu_{ij}(x) = \lim_{(y-x) \rightarrow 0} \frac{p_{ij}(x, y)}{y - x} \quad \text{for } i \neq j$$

$$\mu_{ii}(x) \text{ is defined such that } \sum_j \mu_{ij}(x) = 0$$

$$\text{Hence } \mu_{ii}(x) = \sum_{j \neq i} \mu_{ij}(x) = \lim_{(y-x) \rightarrow 0} \frac{1 - p_{ij}(x)}{y - x}$$

Force of retention

Transition rates: matrix of intensities

$$\boldsymbol{\mu}(\mathbf{x}) = \begin{bmatrix} \mu_{11}(\mathbf{x}) & -\mu_{21}(\mathbf{x}) & \cdot & \cdot & -\mu_{11}(\mathbf{x}) \\ -\mu_{12}(\mathbf{x}) & \mu_{22}(\mathbf{x}) & \cdot & \cdot & -\mu_{12}(\mathbf{x}) \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ -\mu_{11}(\mathbf{x}) & -\mu_{21}(\mathbf{x}) & \cdot & \cdot & \mu_{11}(\mathbf{x}) \end{bmatrix}$$

Discrete-time transition probabilities:

$$\mathbf{P}(\mathbf{x}, y) = \begin{bmatrix} p_{11}(\mathbf{x}, y) & p_{21}(\mathbf{x}, y) & \cdot & \cdot & p_{N1}(\mathbf{x}, y) \\ p_{12}(\mathbf{x}, y) & p_{22}(\mathbf{x}, y) & \cdot & \cdot & p_{N2}(\mathbf{x}, y) \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ p_{1N}(\mathbf{x}, y) & p_{2N}(\mathbf{x}, y) & \cdot & \cdot & p_{NN}(\mathbf{x}, y) \end{bmatrix} \quad \frac{d\mathbf{P}(\mathbf{x})}{dx} = -\boldsymbol{\mu}(\mathbf{x})\mathbf{P}(\mathbf{x})$$

Transition rates: piecewise constant transition intensities (rates)

$$\mathbf{P}(x, y) = \exp[-(y - x)\mathbf{M}(x, y)]$$

$$\exp(\mathbf{A}) = \mathbf{I} + \mathbf{A} + \frac{1}{2!}\mathbf{A}^2 + \frac{1}{3!}\mathbf{A}^3 + \dots$$

$$\exp[-(y - x)\mathbf{M}(x, y)] = \mathbf{I} - (y - x)\mathbf{M}(x, y) + \frac{(y - x)^2}{2!}[\mathbf{M}(x, y)]^2 - \frac{(y - x)^3}{3!}[\mathbf{M}(x, y)]^3 + \dots$$

$$\mathbf{P}(x, y) = \left[\mathbf{I} + \frac{1}{2}\mathbf{M}(x, y)\right]^{-1} \left[\mathbf{I} - \frac{1}{2}\mathbf{M}(x, y)\right]$$

Transition rates: generation and distribution

$$\mu_{ij}(x) = \mu_{i+}(x) \xi_{ij}(x)$$

where $\xi_{ij}(x)$ is the probability that an individual who leaves i selects j as the destination. It is the conditional probability of a *direct transition* from i to j .

Competing risk model

$$\begin{bmatrix} \mu_{11}(x) & -\mu_{21}(x) & \cdot & \cdot & -\mu_{11}(x) \\ -\mu_{12}(x) & \mu_{22}(x) & \cdot & \cdot & -\mu_{12}(x) \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ -\mu_{11}(x) & -\mu_{21}(x) & \cdot & \cdot & \mu_{11}(x) \end{bmatrix} = \begin{bmatrix} \xi_{11}(x) & -\xi_{21}(x) & \cdot & \cdot & -\xi_{11}(x) \\ -\xi_{12}(x) & \xi_{22}(x) & \cdot & \cdot & -\xi_{12}(x) \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ -\xi_{11}(x) & -\xi_{21}(x) & \cdot & \cdot & \xi_{11}(x) \end{bmatrix} \begin{bmatrix} \mu_{1+}(x) & 0 & \cdot & \cdot & 0 \\ 0 & \mu_{2+}(x) & \cdot & \cdot & 0 \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdot & \cdot & \mu_{1+}(x) \end{bmatrix}$$

Transition rates: generation and distribution with covariates

Log-linear model $m_i = \exp[\beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots]$

$$\ln m_i = \beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots$$

Cox model $m_i(x) = m_{i0}(x) \exp[\beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots]$

From transition probabilities to transition rates

The inverse method (Singer and Spilerman)

$$\mathbf{P}(x, y) = \left[\mathbf{I} + \frac{1}{2} \mathbf{M}(x, y) \right]^{-1} \left[\mathbf{I} - \frac{1}{2} \mathbf{M}(x, y) \right]$$

$$\mathbf{M}(x, y) = \frac{y-x}{2} \left[\mathbf{I} - \mathbf{P}(x, y) \right] \left[\mathbf{I} + \mathbf{P}(x, y) \right]^{-1}$$

From 5-year probability to 1-year probability:

$$\mathbf{P}(x, x+1) = \exp\left[-\mathbf{M}(x, x+1)\right]$$

Count data

Poisson model: $\Pr\{N_i = n_i\} = \frac{\lambda_i^{n_i}}{n_i!} \exp[-\lambda_i]$

Covariates: $E[N_i] = \lambda_i = \exp[\beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots]$

$$\ln \lambda_i = \beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots$$

The log-rate model is a log-linear model with an offset:

$$E\left[\frac{N_i}{PY_i}\right] = \frac{\lambda_i}{PY_i} = \exp[\beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots]$$

$$E[N_i] = \lambda_i = PY_i \exp[\beta_{i0} + \beta_{i1}Z_1 + \beta_{i2}Z_2 + \dots]$$

Incomplete data

Poisson model: $\Pr\{N_{ij} = n_{ij}\} = \frac{\lambda_{ij}^{n_{ij}}}{n_{ij}!} \exp[-\lambda_{ij}]$

Data availability: $E[N_{ij}] = \lambda_{ij} = \alpha_i \beta_j$


The *maximization* of the probability is equivalent to maximizing the log-likelihood $l = \sum_{ij} [n_{ij} \ln[\alpha_i \beta_j] - \alpha_i \beta_j]$

$$\hat{\alpha}_i = \frac{n_{i+}}{\sum_j \hat{\beta}_j} \quad \hat{\beta}_j = \frac{n_{+j}}{\sum_i \hat{\alpha}_i}$$

The EM algorithm results in the well-known expression

$$\lambda_{ij} = \frac{n_{i+}}{n_{++}} n_{+j}$$

Conclusion

- Unified perspective on modeling of migration: probability models of counts, probabilities (proportions) or rates (*risk indicators*)
 - State occupancies and state transitions
 - Transition rate = exit rate * destination probabilities
- 
Timing of event Direction of change