

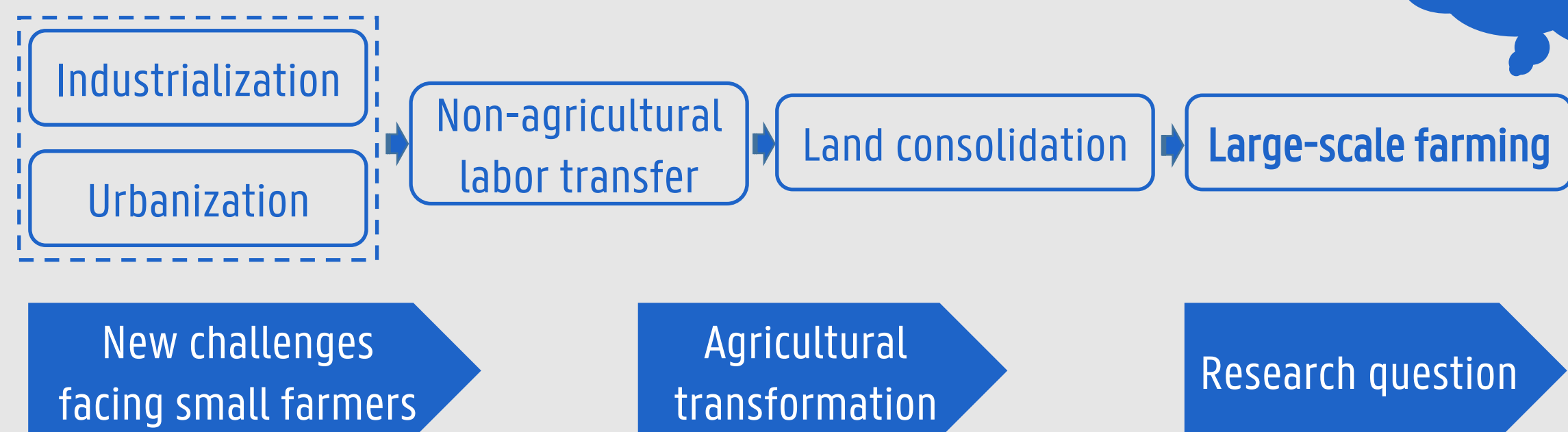
RURAL DEVELOPMENT ECONOMICS

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EXPANDING FARMS: A SPATIAL PANEL DATA ANALYSIS TO EXPLORE THE DEVELOPMENT OF LARGE-SCALE FARMING IN CHINA

Introduction

Background



Objectives

- Examining the determinants that influence the development of large-scale farming while taking into account spatial interrelationship;
- Proposing policy suggestions on how to promote the sound development of large-scale farming.

Methodology

Data collection

Statistical data covers 44 county-level cities and counties of Jiangsu Province in China from 2002 to 2016.

Model specification

(1) Spatial Autoregressive Model (SAR)

$$y = \rho(I_T \otimes W_N)y + X'\beta + \varepsilon$$

(2) Spatial Error Model (SEM)

$$y = X'\beta + \mu$$

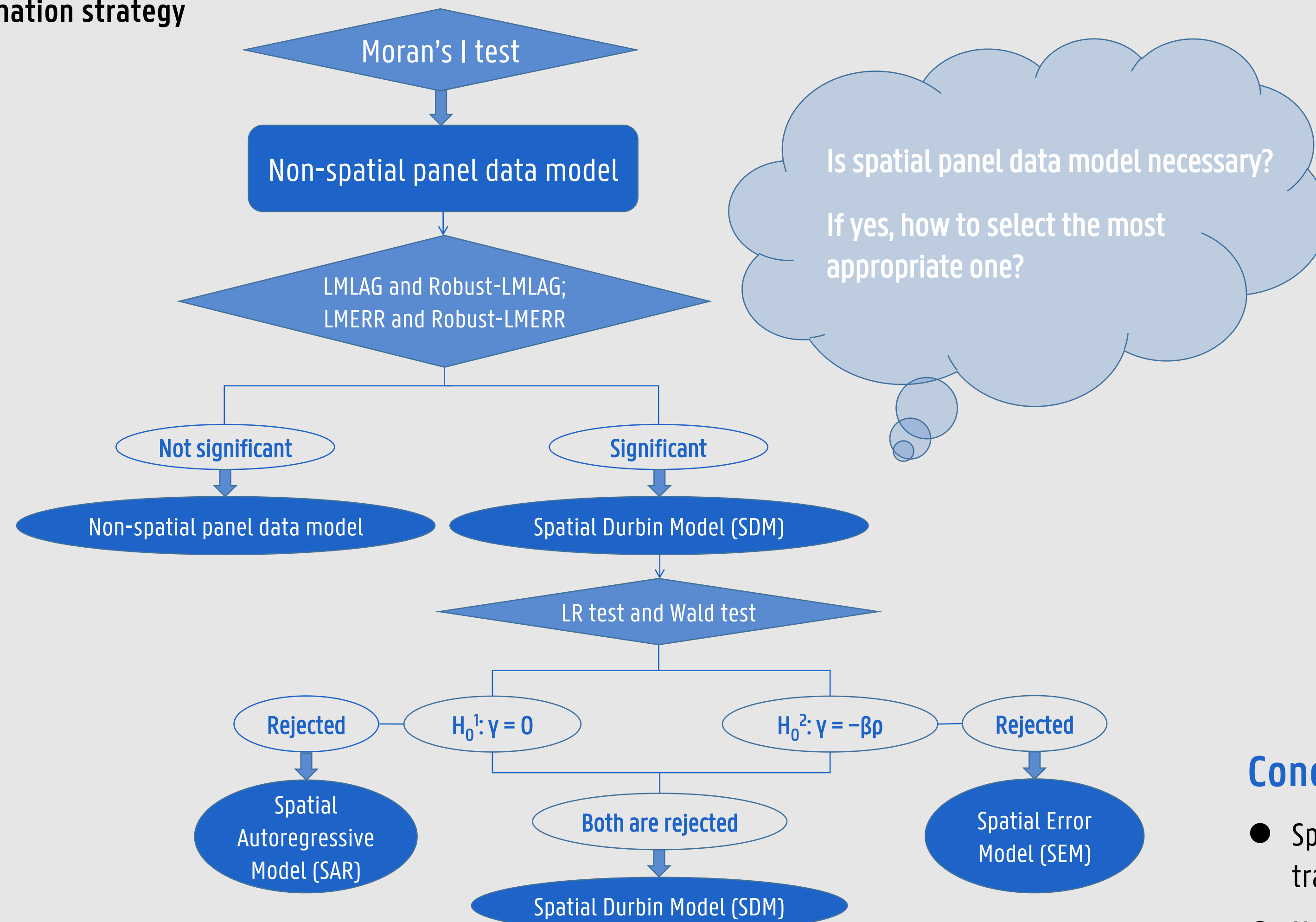
$$\mu = \lambda(I_T \otimes W_N)\mu + \varepsilon$$

(3) Spatial Durbin Model (SDM)

$$y = \rho(I_T \otimes W_N)y + X'\beta + \gamma(I_T \otimes W_N)X + \varepsilon$$

y is dependent variable; X is the matrix of independent variables with its corresponding matrix of coefficients β ; ε is the error term; W_N is a $(N \times N)$ spatial weighting matrix indicates the interaction between any two counties; ρ is the spatial autoregressive coefficient; μ reflects the spatially autocorrelated error term; λ represents the spatial autocorrelation coefficient of the error term; γ is the spatial autocorrelation coefficient of independent variables.

Estimation strategy



Results

Spatial weight matrix - Inverse distance weight matrix: Threshold distance

Table 1 Randomization results of Moran's I-statistic

| Critical distance cut-off (km) | Moran's I (Standardized Z-value) | | |
|--------------------------------|----------------------------------|---------------------------|---------------------------|
| | lnSCALE _{t=2002} | lnSCALE _{t=2008} | lnSCALE _{t=2016} |
| min | 0.315***(2.365) | 0.246**(1.983) | 0.417***(3.065) |
| 60 | 0.335***(3.163) | 0.343***(3.289) | 0.414***(3.903) |
| 70 | 0.310***(3.294) | 0.218***(2.512) | 0.304***(3.237) |
| 80 | 0.297***(3.573) | 0.172**(2.185) | 0.263***(3.179) |
| 90 | 0.288***(3.809) | 0.142**(2.050) | 0.259***(3.424) |
| 100 | 0.314***(5.136) | 0.048(1.066) | 0.214***(3.461) |

Note: SCALE=Land area cultivated under large-scale farming/Total cultivated land in the county; *** p<0.01, ** p<0.05, * p<0.1



Moran's I-statistic of the dependent variable

- Moran's I values are positive at 1% significance level during the whole period;
- Implying that the promotion of large-scale farming in Jiangsu Province has a significant positive correlation in spatial distribution.

Estimates of non-spatial panel data model

- LR test: The model with two-way fixed effects is justified as the best fitting for the non-spatial panel data specification.
- LM tests and their robust counterparts: Most of the null hypothesis of no spatially lagged dependent variable and no spatially autocorrelated error term are strongly rejected at 1% significance level in all the specifications.

➔ There exists spatial dependence among the data; Spatial panel model is necessary for further analysis.

Estimates of spatial panel data model

- LR test and Wald test: SDM model is more appropriate than SAR model and SEM model.
- Hausman test: Fixed effects assumption provides a better fit to the given data.

➔ SDM model with spatial and time-period fixed effects is chosen as the best specification.

Model results: The estimated coefficient on the spatially lagged term ρ of the dependent variable is significantly positive at 1% level, suggesting that the development of large-scale farming in neighboring counties exerts a positive effect on local large-scale farm expansion.

- However, the interpretation of spatial spillovers should not stop at the point estimates above; a better interpretation lies in the estimation of indirect effects presented below.

Table 2 Cumulative impacts from SDM model with spatial and time-period fixed effects

| Variables | Direct effect | Indirect effect | Total effect |
|--------------------------------------|-------------------------|-----------------------|-------------------------|
| Total land area | -1.7865*** (-6.1371) | -0.2576 (-0.5801) | -2.0442*** (-3.8818) |
| Land area per household | -3.1014* (-1.6981) | 0.1446 (0.0318) | -2.9568 (-0.5723) |
| Land area per household ² | 1.2424** (2.1648) | -0.2434 (-0.1772) | 0.9990 (0.6432) |
| GDP per capita | 0.2829 (1.5919) | 0.1372 (0.4299) | 0.4201 (1.1172) |
| Share of tertiary industry | -0.7216 (-0.7531) | 2.1367 (1.5718) | 1.4152 (0.8812) |
| Non-agricultural employment | 1.5383** (2.4690) | 5.2153*** (5.5958) | 6.7536*** (6.7038) |
| Agricultural mechanization | 0.1619 (0.4678) | 1.3760** (2.4706) | 1.5379*** (2.7540) |
| Land certification | 0.6179*** (4.2266) | -0.0975 (-0.4384) | 0.5204* (1.9136) |

Note: z-statistics in parentheses; *** p<0.01, ** p<0.05, * p<0.1

Conclusions

- Spatial spillovers do exist in the development of large-scale farming between counties in Jiangsu Province; traditional non-spatial panel data model will lead to biased results due to model misspecification;
- Non-agricultural employment, agricultural mechanization as well as land certificate issuance have been found to have positive effects on the promotion of large-scale farming;
- Policy-makers should target their policy for large-scale farm expansion in an overall view, considering not only the development situation of local regions but also possible influence on surrounding regions.

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