

Healthcare Accessibility and Medical Expenditure: Evidence from Rural China

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Introduction

Unequal Distribution of Healthcare Resources Between Urban and Rural Areas:

- China's healthcare system has expanded rapidly in recent years (Li Fu, 2017; Lancet, 2019), but **urban-rural disparities remain significant** (Yip et al., 2012; Meng et al., 2019).
- Rural areas face persistent shortages of medical resources (Li et al., 2017; Sun et al., 2021):
 - Insufficient healthcare facilities
 - Fewer qualified medical professionals
 - Poor accessibility to basic services
- Inadequate access to healthcare can lead to (Yu et al., 2020; Zhang et al., 2023):
 - Delayed treatment
 - Increased health risks
 - Higher medical expenditure

Research Question:

- Do rural residents face higher medical costs due to limited access to healthcare services?

Introduction

Empirical Challenges:

- Existing studies mainly focus on **urban populations or national-level data**, with limited attention to rural dynamics over time (Li et al., 2019; Xiong et al., 2012).
- There is a lack of **long-term, micro-level empirical evidence** on how healthcare accessibility affects rural medical expenditure (Yu et al., 2020).

This study utilizes panel data from the China Rural Fixed Observation Point System (2009–2017)—

- Covers 356 rural sites, 20,745 households, and 488,234 individual records.
- Offers rich and continuous data on income, health, and medical use.

Objective:

- Investigate how accessibility to healthcare resources influences the medical spending of rural residents.

Key Empirical Findings

1. **Significant negative correlation** between distance to medical institutions and medical expenditure: A 1% increase in distance leads to a 0.1–0.3% decrease in average medical spending.
2. **Heterogeneity analysis:** The suppressing effect of distance is more pronounced among:
 - Suburban residents
 - Individuals with poor health
 - Low-income households
3. **Mechanism analysis:** Greater distance reduces the frequency of doctor visits and total treatment costs, which ultimately lowers medical expenditure.
4. **Extended result:** Increased distance to healthcare facilities also has a **significant negative impact on long-term health status**.

Data & Variables

China Rural Fixed Observation Point System (2009–2017)

- Covers **356 rural observation points**, **20,745 households**, and **488,234 individual-level records**.
- Provides panel data on: Household income and expenditure

Amap Point-of-Interest (POI) Data

- *Healthcare facilities*: Tertiary Grade-A hospitals Primary healthcare institutions (village clinics, outpatient clinics)
- *Government institutions*: Sub-township, township, prefecture, provincial, and national levels

Linkage: Datasets are matched geographically to measure healthcare resource accessibility.

Research Design

Empirical Specification:

$$Outcome_{ihct} = \alpha + \beta dist_{ct} + X'_{ict}\delta + \mu_i + \gamma_t + \varepsilon_{ict} \quad (1)$$

Where:

- y_{ihct} : Average medical expenditure of individual i in household h , in county c , year t .
- β : Measures the effect of distance to healthcare facilities on medical expenditure
- $dist_{ct}$: Distance from village c to the nearest healthcare facility in year t
- X_{ict} : Vector of control variables at the individual and household level.
- μ_i : Individual fixed effects
- γ_t : Year fixed effects
- Standard errors are clustered at the village level.

Two-Stage Regression Model

First Stage:

$$Z_{ct} = \alpha + \beta \text{dist}_{ct} + X'_{ict} \delta + \mu_c + \gamma_t + \varepsilon_{ict}$$

Second Stage:

$$y_{ihct} = \lambda_0 + \lambda_1 \widehat{Z}_{ct} + X'_{ict} \delta + \mu_c + \gamma_t + \nu_{ict}$$

- y_{ihct} : Average medical expenditure of individual i in household h , county c , year t
- dist_{ct} : Distance from village c to the nearest tertiary hospital
- Z_{ct} : Distance to the nearest actual healthcare facility
- \widehat{Z}_{ct} : Predicted value of Z_{ct} from the first stage (instrumented)
- X_{ict} : Vector of control variables (e.g., income, household size, health status)
- μ_c, γ_t : County and year fixed effects
- $\varepsilon_{ict}, \nu_{ict}$: Error terms

Basic Outcomes

| VARIABLES | (1) | (2) | (3) | (4) |
|---|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| | | Avg. Medical Cost | | |
| <u>log_tertiary</u> | -0.147** (0.0589) | -0.150** (0.0587) | -0.146** (0.0553) | -0.147** (0.0557) |
| Household-level characteristics (e.g., gender, income, land, labor) | Y | Y | Y | Y |
| Controls for topographic features (e.g., terrain type, urban/suburban classification) | N | Y | Y | Y |
| Controls for county-level characteristics (e.g., gov. seat, economic dev.) | N | N | Y | Y |
| Controls for village-level demographics & economy | N | N | N | Y |
| Constant | 2.325*** (0.556) | 2.022*** (0.645) | 2.237*** (0.522) | 2.082*** (0.564) |
| Observations | 488,234 | 485,494 | 485,494 | 485,494 |
| R-squared | 0.014 | 0.014 | 0.015 | 0.016 |
| Number of individual_id | 89,931 | 89,820 | 89,820 | 89,820 |
| Age Group | Y | Y | Y | Y |
| Health Status | Y | Y | Y | Y |
| Primary Income Source | Y | Y | Y | Y |
| Primary Household Occupation | Y | Y | Y | Y |
| Occupation | Y | Y | Y | Y |
| Individual Fixed Effects | Y | Y | Y | Y |
| Time Fixed Effects | Y | Y | Y | Y |

2SLS Outcomes

| VARIABLES | Tertiary Grade-A Hospital Avg. Medical Cost | Primary Healthcare |
|---|--|------------------------------|
| <u>log_distance</u> | -0.390*** (0.0486) | -0.414*** (0.0517) |
| Household-level characteristics (e.g., gender, income, land, labor) | Y | Y |
| Controls for topographic features (e.g., terrain type, urban/suburban classification) | Y | Y |
| Controls for county-level characteristics (e.g., gov. seat, economic dev.) | Y | Y |
| Controls for village-level demographics & economy | Y | Y |
| Other control <u>variables</u> | Y | Y |
| Observations | 477,842 | 477,842 |
| R-squared | 0.014 | 0.013 |
| Number of individual_id | 82,168 | 82,168 |
| Age Group | Y | Y |
| Health Status | Y | Y |
| Individual Fixed Effects | Y | Y |
| Time Fixed Effects | Y | Y |

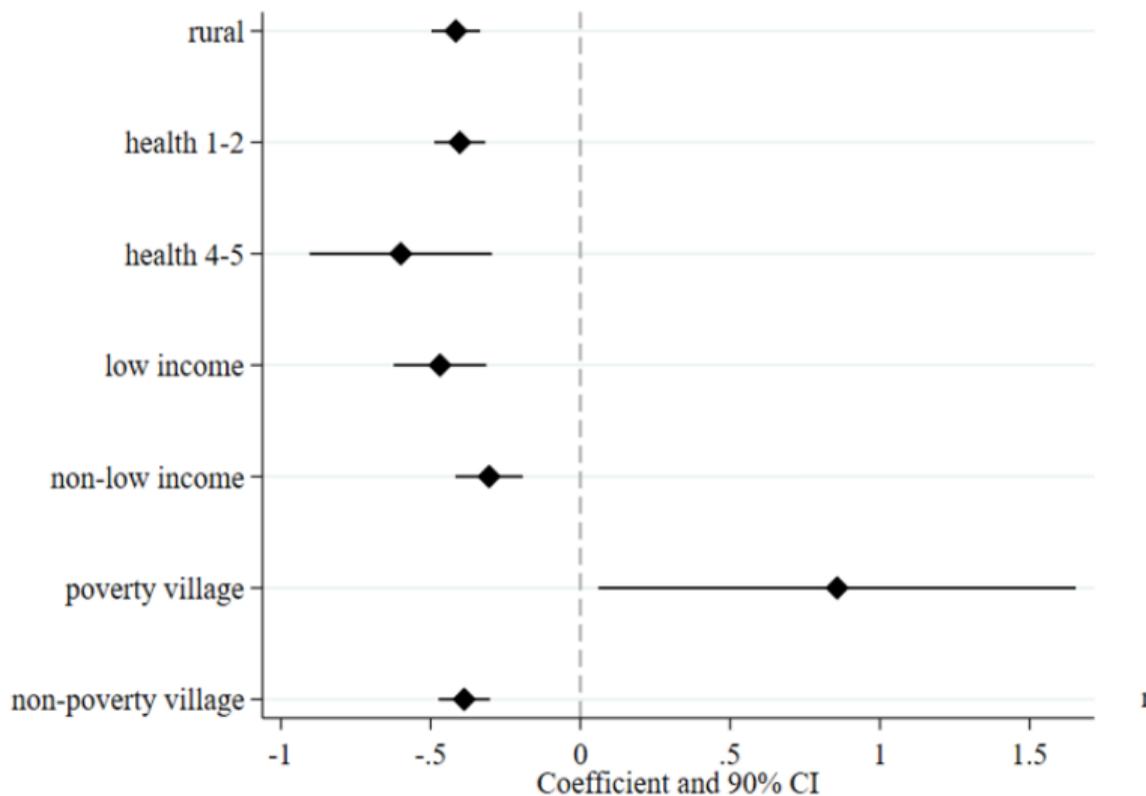
Does Distance to Healthcare Affect Medical Visit Decisions?

| VARIABLES | (1) | (2) |
|--------------------------|-------------------------------------|-------------------------------------|
| | Medical Visit (Yes/No) | |
| log_tertiary | -0.0307** (0.0114) | |
| <u>log_primary</u> | | -0.0263** (0.0114) |
| Constant | 0.403*** (0.115) | 0.374*** (0.113) |
| Other control variables | Y | Y |
| Observations | 485,494 | 485,494 |
| R-squared | 0.012 | 0.011 |
| Number of individual_id | 89,820 | 89,820 |
| Age Group | Y | Y |
| Health Status | Y | Y |
| Individual Fixed Effects | Y | Y |
| Time Fixed Effects | Y | Y |

Suburban vs. Rural

| VARIABLES | 2SLS Suburban Area | 2SLS Rural Area |
|--------------------------------|-----------------------|-----------------------|
| <u>log_tertiary</u> | -12.55** (5.840) | -0.416*** (0.0494) |
| Other control <u>variables</u> | Y | Y |
| Observations | 85,020 | 388,253 |
| R-squared | -3.902 | 0.013 |
| Number of individual_id | 18,140 | 67,924 |
| Age Group | Y | Y |
| Health Status | Y | Y |
| Individual Fixed Effects | Y | Y |
| Time Fixed Effects | Y | Y |

Heterogeneity Analysis



Health Status Change in the Past Year

| VARIABLES | (1) | (2) |
|--------------------------|---------------------------------------|------------------------------|
| | Health Status Change in the Past Year | |
| log_tertiary | 0.0393*** (0.0117) | |
| log_primary | | 0.0448*** (0.0134) |
| Other control variables | Y | Y |
| Observations | 388,751 | 388,751 |
| R-squared | 0.043 | 0.043 |
| Number of individual_id | 74,703 | 74,703 |
| Age Group | Y | Y |
| Health Status | Y | Y |
| Individual Fixed Effects | Y | Y |
| Time Fixed Effects | Y | Y |

Robustness Checks

| VARIABLES | Tertiary Grade-A Hospital | Primary Healthcare Institution | Healthcare Facility |
|---|---------------------------|--------------------------------|---------------------|
| | Avg. Medical Cost | | |
| log_tertiary | -0.390*** | | |
| | -0.0486 | | |
| log_primary | | -0.414*** | |
| | | -0.0517 | |
| log_medical | | | -0.390*** |
| | | | (0.0486) |
| Household-level characteristics (e.g., gender, income, land, labor) | Y | Y | Y |
| Controls for topographic features (e.g., terrain type, urban/suburban classification) | Y | Y | Y |
| Controls for county-level characteristics (e.g., gov. seat, economic dev.) | Y | Y | Y |
| Controls for village-level demographics & economy | Y | Y | Y |
| other control variables | Y | Y | Y |
| Observations | 477,842 | 477,842 | 477,842 |
| R-squared | 0.014 | 0.013 | 0.014 |
| Number of individual_id | 82,168 | 82,168 | 82,168 |
| Age Group | Y | Y | Y |
| Health Status | Y | Y | Y |
| Individual Fixed Effects | Y | Y | Y |
| Time Fixed Effects | Y | Y | Y |

Conclusion

- Limited accessibility to healthcare significantly reduces medical utilization and expenditures in rural areas.
- Vulnerable populations—such as the poor, the unhealthy, and those in suburban areas—are disproportionately affected by distance barriers.
- While reduced expenditure may appear favorable, it actually reflects **underutilization of necessary medical care**, potentially leading to worse long-term health outcomes.
- Improving healthcare accessibility is essential for promoting **health equity and long-term welfare** in rural China.

Thank You!