The Illness-Poverty-Amenity Trap: Evidence from 7 Million Seniors

Jonathan D. Ketcham*           Nicolai V. Kuminoff†           Sophie Mathes◊

* Arizona State University Marketing Department
† Arizona State University Economics Department, and NBER
◊ Arizona State University Economics Department

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The Illness-Poverty-Amenity Trap
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Diagram:
- Location
- Pollution
- Income
- Health

Relations:
- Location determines exposure to Pollution
- Sorting
- Out of pocket cost
- Health
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Studying Seniors is Important

• 20% of the U.S. population by 2030

• Annual taxpayer-financed Medicare spending >$600 billion

• Most vulnerable age group to pollution and heat stress

  e.g. U.S. EPA attributes 78% of all premature deaths avoided by air pollution regulations to people 65+
Connecting three literatures

1. **Intergenerational poverty trap**
   Becker and Tomes (JPE 1979), Loury (ECMA 1981), Benabou (EER 1994), Galor and Ziera (REStud 1993), Durlauf (JEG 1996), Azariadis and Stachurski (HEG 2005), Durlauf and Seshadri (NBER 2018), Chetty et al. (QJE 2018)

2. **Residential sorting based on preferences and income**
   Tiebout (JPE 1956), Epple and Platt (JUE 1998), Epple and Sieg (JPE 1999), Sieg et al. (IER 2004), Smith et al. (JEEM 2004), Banzhaf and Walsh (AER 2007), Bayer, Keohane and Timmins (JEEM 2009), Kuminoff (JEEM 2009), Kuminoff, Smith and Timmins (JEL 2013), Hamilton and Phaneuf (JEEM 2015), Depro, Timmins and O’Neil (JAERE 2015)

3. **Health impacts of local pollution, climate, and health care**
   Pope et al. (JAMA 2002), Graff-Zivin and Neidell (JEL 2013), Underwood (Science 2017), Bishop, Ketcham and Kuminoff (NBER 2018), Deryugina and Molitor (NBER 2018), Zhang et al. (PNAS 2018), Finkelstein, Gentzkow and Williams (2019)
This Research

We develop a conceptual model of residential sorting, pollution exposure and health and use it to define mechanisms that could generate an illness-poverty-amenity trap. Then we test for the presence of those mechanisms using Medicare administrative records on more than 7 million U.S. seniors from 2001-2013, finding evidence of an IPA trap.
Preferences and health

Utility for retiree $i$ at age $t$ in location $j$: \[ U_{ijt} = u(b, q, g_{jt}; h_{it}), \] where

- $b$ = numeraire private good
- $q$ = housing quantity
- $g_{jt}$ = local amenities (e.g. air pollution, climate, health care quality)
- $h_{it}$ = stock of health

Evolution of the health stock: \[ h_{it+1} = f(h_{it}, g_{jt}, m_{it}, t, \varepsilon_{it}), \] where

- $m_{it}$ = medical expenditures
- $\varepsilon_{it}$ = idiosyncratic health shock
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Budget constraint:  
\[ y_i - m(h_{it}) = \hat{y}_i = b - p_j q \]

- $y_i$ = permanent income (e.g. pension, social security)
- $m(h_{it})$ = medical expenditures determined by health
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Repeated Static Optimization Problem

\[
\max_{j,b,q} U(b, q, g_{jt}; h_{it})
\]

Budget constraint: 
\[
y_i - m(h_{it}) = \hat{y}_i = b - p_j q
\]

Evolution of the health stock: 
\[
h_{it} = f(h_{it-1}, g_{jt-1}, m_{it-1}, t - 1, \varepsilon_{it-1})
\]

Note: we follow the vast majority of residential sorting literature in abstracting from forward-looking behavior; e.g. w.r.t. health.
Indirect utility & single-crossing

Indirect utility: \( V(g, p, \hat{y}, h) = U[y - p \cdot q(g, p, \hat{y}, h), q(g, p, \hat{y}, h), g, h] \)

Indirect indifference curve: \( M(g, p, \hat{y}, h) = -\frac{\partial V(g, p, \hat{y}, h)}{\partial g} \frac{\partial g}{\partial V(g, p, \hat{y}, h)} \)

Single crossing condition: \( M(\cdot) \) is strictly increasing in \((\hat{y}|h)\) and \((h|\hat{y})\)

\( \Rightarrow \) Sorting equilibrium will reflect “stratification” by income and health with “ascending bundles” in price-amenity space.

Proof: Epple and Sieg (JPE 1999)
Stratification by health and income

\[ \text{wlog: } g_1 < g_2 < g_3 \]

\[ \text{SCC } \Rightarrow p_1 < p_2 < p_3 \]
Stratification by health and income

wlog: $g_1 < g_2 < g_3$

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\[ w \log: \quad g_1 < g_2 < g_3 \]

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Health dynamics and the IPA trap

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\[ \hat{y}_i = y_i - m(h_{it}) \]

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\[ h_{it} = f(h_{it-1}, g_{jt-1}, m_{it-1}, t, \varepsilon_{it-1}) \]

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Testable Hypotheses

1. Lower income seniors will tend to:
   A. live in lower amenity areas
   B. become sicker sooner
   C. spend more on health care

2. Health shocks will tend to:
   A. increase medical spending
   B. increase migration

3. Low income migrants will tend to move to low amenity places
Data: Medicare administrative records

A random 10% sample of all Medicare A,B enrollees from 2001-2013 (over 7 million people over age 65)

• **Demographics**: race, gender, birth date, death date, Medicare expenditures (gross and out-of-pocket [OOP]), Medicaid subsidy (2006-2013), state buy-in Medicaid proxy (2001-2013)

• **First diagnosis of 30+ chronic medical conditions**: acute myocardial infarction, anxiety, asthma, atrial fibrillation, bipolar disorder, breast cancer, cataract, chronic kidney disease, chronic obstructive pulmonary disease, colorectal cancer, congestive heart failure, dementia, depression, diabetes, endometrial cancer, fibromyalgia, glaucoma, hip fracture, hyperlipidemia, hypertension, hypothyroidism, ischemic heart disease, lung cancer, obesity, osteoporosis, peripheral vascular disease, prostate cancer, rheumatoid arthritis, schizophrenia, stroke

• **Residential ZIP+4 codes**: Census block group data on education, income, housing stock and neighborhood demographics
Residential locations defined by ZIP+4 centroids

- Over 35 million residential ZIP+4 codes in U.S., 1 for every 3.3 households
- **Examples**: apartment building floor, one side of one street on a city block
EPA’s Air Quality Monitoring Network for PM$_{2.5}$
Inferring Exposures: ZIP+4 w/ Inverse Distance Weighting
H1.A: Low income seniors live in low amenity areas

• Low-income group has 5% higher exposure to PM$_{2.5}$ in 2001
H1.A: Low income seniors live in low amenity areas

- PM$_{2.5}$ exposure gap narrowed after EPA strengthened regulations
## Summary statistics

<table>
<thead>
<tr>
<th></th>
<th>Traditional Medicare (A,B)</th>
</tr>
</thead>
<tbody>
<tr>
<td>number of individuals</td>
<td>7,356,473</td>
</tr>
<tr>
<td>mean age at entry</td>
<td>71</td>
</tr>
<tr>
<td>number of years in the sample</td>
<td>8</td>
</tr>
<tr>
<td>ever moved (%)</td>
<td>18</td>
</tr>
<tr>
<td>died before December 31, 2013</td>
<td>41</td>
</tr>
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</table>

- 2010 Medicare spending per capita = $11,852
  - Taxpayer subsidy = $10,044
  - Out-of-pocket spending = $1,728
  - OOP spending is 7% of seniors’ median income ($25,757)
H1.B: Low income seniors become sicker faster

- Cohort of 269,056 people who were 66 years old in 2001 (10.6% low income)
- State buy-in used as a proxy for Medicaid status (corr > 0.9 after 2006)
H1.B: Low income seniors become sicker faster

- Low-income groups have higher mortality within every age cohort
H1.C: Low income seniors spend more on health care

- Annual average OOP differential ranges from $590 to $760
H2.A: Health shocks increase medical spending

\[ \Delta m_{ijt} = \beta_t \Delta h_{ijt-1} + \gamma_t x_{ijt-1} + \omega_{jt-1} + \epsilon_{ijt}, \]  

where

\[ \Delta m_{ijt} = \text{annual change in out-of-pocket medical spending} \]
\[ \Delta h_{ijt-1} = \text{new chronic condition diagnoses} \]
\[ x_{ijt-1} = \text{integer age x gender dummies, race dummies} \]
\[ \omega_{jt-1} = \text{residential county in year } t-1 \]
\[ \epsilon_{ijt} = \text{orthogonal error} \]

- Regression coefficients and fixed effects evolve flexibly over time
- Identification comes from within-person health shocks.
H2.A: Health shocks increase medical spending

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acute myocardial infarction
dementia
lung cancer

anxiety
depression
obesity

asthma
diabetes
osteoporosis

atrial fibrillation
endometrial cancer
peripheral vascular disease

bipolar disorder
fibromyalgia
prostate cancer rheumatoid arthritis

breast cancer
glomerulonephritis
arthritis schizophrenia

cataract
hip fracture
stroke

cataract
hyperlipidemia

chronic kidney disease
hypertension

chronic obstructive pulmonary disease
hypothyroidism

colorectal cancer
ischemic heart disease

congestive heart failure

H2.A: Health shocks increase medical spending
H2.A: Health shocks increase medical spending

- Point estimates and 95% CI’s from annual regressions of 3 to 4 million people
- In 2010, median income among people 65+ was $25,757
H2.A: Health shocks increase medical spending

- Large effects for cancer and diseases that impair mobility

| Average effect on OOP spending from 2002 to 2013 ($) |
|-----------------|----------------------------------|
| 2,809           | hip fracture                     |
| 2,679           | lung cancer                      |
| 2,216           | colorectal cancer                |
| 2,056           | breast cancer                    |
| 1,918           | endometrial cancer               |
| 1,712           | prostate cancer                  |
| 1,279           | stroke                           |
| 1,250           | heart attack                     |
| 1,095           | Alzheimer’s disease and related dementias |
| 731             | ischemic heart disease           |

- Small effects for vision impairments (cataract, glaucoma), and conditions treated with generic drugs (hypertension, hyperlipidemia)
H2.A: Health shocks increase medical spending

- Large effects for cancer and diseases that impair mobility

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Linked to PM$_{2.5}$ exposure by economic & epi/med studies

- Small effects for vision impairments (cataract, glaucoma), and conditions treated with generic drugs (hypertension, hyperlipidemia)
H2.B: Health shocks increase migration

\[ \{ j \neq j - 1 \} = \beta_t \Delta h_{ijt-1} + \gamma_t x_{ijt-1} + \omega_{jt-1} + \varepsilon_{ijt}, \text{ where} \]

\{ j \neq j - 1 \} = 1 \text{ iff person } i \text{ moved in year } t \\
\Delta h_{ijt-1} = \text{ new chronic condition diagnoses} \\
x_{ijt-1} = \text{ age x gender dummies, race, Medicaid eligibility} \\
\omega_{jt-1} = \text{ residential county in year } t-1 \\
\varepsilon_{ijt} = \text{ orthogonal error} \\

• Regression coefficients and fixed effects evolve flexibly over time \\
• Identification comes from within-person health shocks
H2.B: Health shocks increase migration

- Dementia
- Hip fracture
- Hypertension
H2.B: Health shocks increase migration
H2.B: Health shocks increase migration

- Largest effects for diseases that impair cognition and/or mobility

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- Null effects for cancers, vision impairments (cataract, glaucoma), and conditions with mild symptoms (hypertension, hyperlipidemia)
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Linked to PM$_{2.5}$ exposure by economic & epi/med studies

- Null effects for cancers, vision impairments (cataract, glaucoma), and conditions with mild symptoms (hypertension, hyperlipidemia)
H3: Low income migrants go to lower amenity areas

\[ \Delta PM_{it} = \alpha PM_{it-1} + \gamma \{ y_{it} \leq y^* \} \{ age_{2001} \} + \delta x_{ijt} + \epsilon_{ijt}, \]

where \( \Delta PM_{it} = \) change annual average residential PM$_{2.5}$ exposure

\( \{ j \neq j - 1 \} = 1 \) iff person $i$ moved in year $t$

\( \{ y_{it} \leq y^* \} = \) low-income indicator (based on state buy-in)

\( x_{ijt-1} = \) age dummies, year dummies

\( \epsilon_{ijt} = \) orthogonal error

- Identified by differences in pollution at destination locations among high and low income groups of the same age who emigrated from similarly polluted areas
H3: Low income migrants go to lower amenity areas

- Average move increases the income-pollution gap by 0.09 μg/m³
H3: Low income migrants go to lower amenity areas

- Average differential of 0.09 μg/m³ is equivalent to 17% of the pollution exposure gap that existed in 2001
Summary

• We extended the residential sorting literature to consider interactions between location choice, pollution exposure, illness and poverty

• Consistent with model predictions, we found that:
  o Lower income seniors live in more polluted areas, spend more on health care, and die sooner
  o Health shocks substantially increase medical spending and patient migration
  o Lower income migrants move to relatively dirtier areas

• Future research: structural model of “Tiebout (JPE 1956)-Grossman (JPE 1972)” interactions between residential location, pollution exposure, and health