

Energy Insecurity in Redlined America

AERE 2021

A. Justin Kirkpatrick
Michigan State University
June 2021

Generous funding provided by the Nicholas Institute at Duke

[Most recent web-based slides here](#)

Energy Insecurity

The disproportionate share of household income allocated to energy expenses with those that exceed a 10% threshold categorized as experiencing "energy insecurity." (Hernández 2015)

- Drehabl and Ross (2016) find 75th percentile energy burdens above 26%.
- Lyubich (2020) finds minority households spend more on energy
- Doremus et al (2021) finds low-income households use more energy during weather extremes

Energy Insecurity

The disproportionate share of household income allocated to energy expenses with those that exceed a 10% threshold categorized as experiencing "energy insecurity." (Hernández 2015)

- Drehabl and Ross (2016) find 75th percentile energy burdens above 26%.
- Lyubich (2020) finds minority households spend more on energy
- Doremus et al (2021) finds low-income households use more energy during weather extremes

Energy Inequity

- Reames (2016): Minority-dominated census block-groups tend to have lower (worse) energy efficiency and spend a greater total amount **for the same level of energy services relative to non-minority households.**
- Drehabl and Ross (2016) using ACS data: Black and Hispanic households face higher median energy burdens, even conditional on income.

Energy Insecurity

The disproportionate share of household income allocated to energy expenses with those that exceed a 10% threshold categorized as experiencing "energy insecurity." (Hernández 2015)

- Drehabl and Ross (2016) find 75th percentile energy burdens above 26%.
- Lyubich (2020) finds minority households spend more on energy
- Doremus et al (2021) finds low-income households use more energy during weather extremes

Energy Inequity

- Reames (2016): Minority-dominated census block-groups tend to have lower (worse) energy efficiency and spend a greater total amount **for the same level of energy services relative to non-minority households.**
- Drehabl and Ross (2016) using ACS data: Black and Hispanic households face higher median energy burdens, even conditional on income.

Energy Inequity: "The disproportionate incidence of energy insecurity in heavily-minority areas relative to non-minority areas of similar income."

Why?

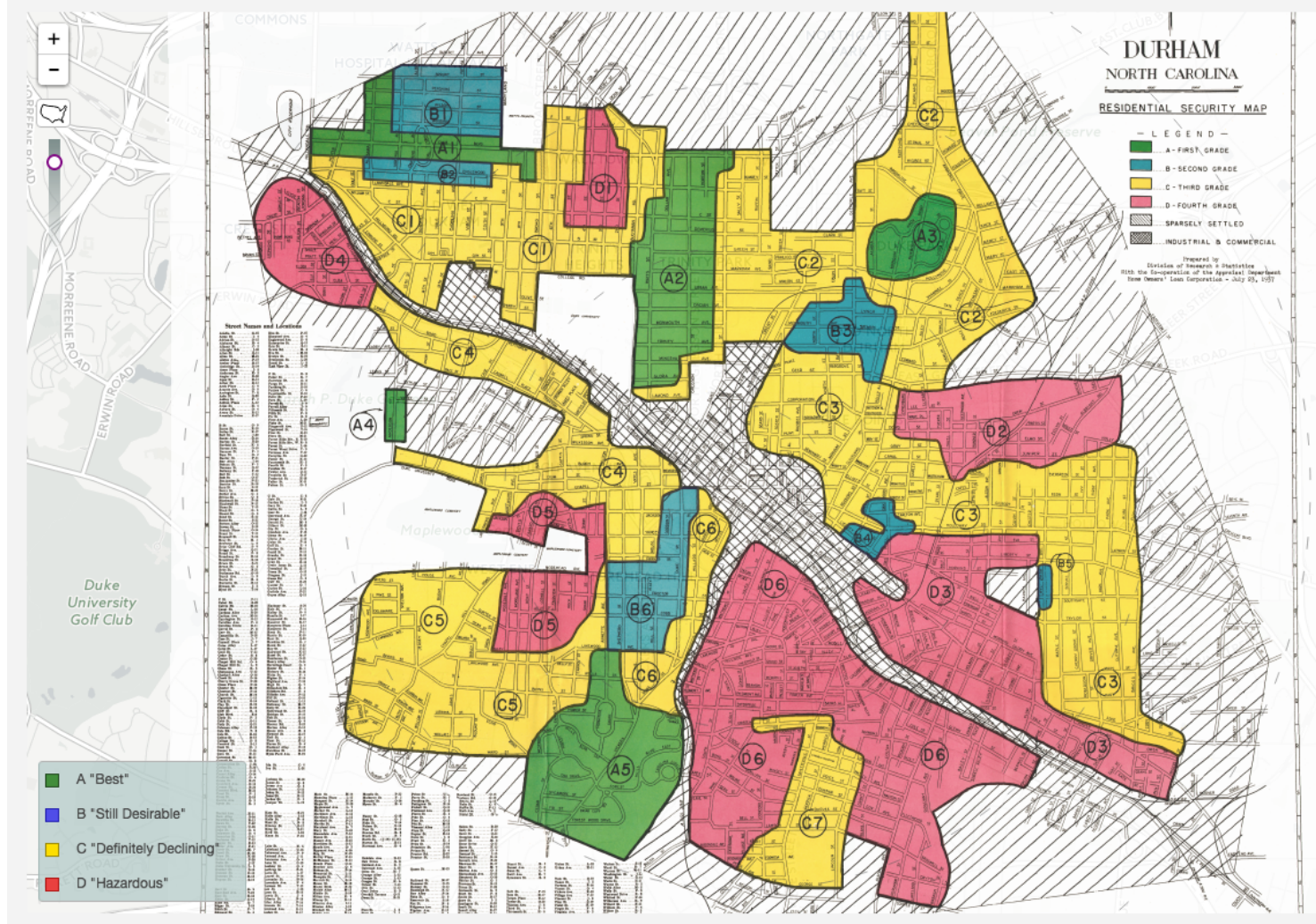
- Preferences & Sorting?
 - Lower-efficiency homes are less expensive, income constraints → "coming to the nuisance" (Banzhaf, 2011)
 - But conditional on income, do minority households prefer lower efficiency?
 - Lower utility of heating? Lower utility of non-energy consumption?

Why?

- Preferences & Sorting?
 - Lower-efficiency homes are less expensive, income constraints → "coming to the nuisance" (Banzhaf, 2011)
 - But conditional on income, do minority households prefer lower efficiency?
 - Lower utility of heating? Lower utility of non-energy consumption?
- Current housing discrimination or heterogeneous information?
 - Christensen et al (2020): Rental agents steer minority households away from low-toxic exposure properties
 - Missing information disproportionately affects those sorting to the lower quality homes (Bakkensen and Ma, 2020; Hausman and Stolper, 2020)

Why?

- Preferences & Sorting?
 - Lower-efficiency homes are less expensive, income constraints → "coming to the nuisance" (Banzhaf, 2011)
 - But conditional on income, do minority households prefer lower efficiency?
 - Lower utility of heating? Lower utility of non-energy consumption?
- Current housing discrimination or heterogeneous information?
 - Christensen et al (2020): Rental agents steer minority households away from low-toxic exposure properties
 - Missing information disproportionately affects those sorting to the lower quality homes (Bakkensen and Ma, 2020; Hausman and Stolper, 2020)
- State dependence / hysteresis
 - Historic forms of discrimination
 - Frictions in moving costs



Durham, NC Redlining Map (source: URichmond Mapping Inequality)

Homeowners Loan Corporation (HOLC)

- New Deal agency tasked with assessing mortgage risk for federal refinancing efforts
- Neighborhoods risk-graded by local agents 1933-1939
- Largely considered "subversive minorities" to be harbinger of decline and risk.
- *De facto* discrimination in housing was widespread

NS FORM-B
2-3-37

AREA DESCRIPTION
(For Instructions see Reverse Side)

1. NAME OF CITY Durham, N. C. SECURITY GRADE C AREA NO. 5

2. DESCRIPTION OF TERRAIN Rolling

3. FAVORABLE INFLUENCES: All city conveniences, fair transportation

4. DETRIMENTAL INFLUENCES: Cemetery on north, and old amusement park

5. INHABITANTS: Mechanics, tobacco workers,
a. Type Clerks; b. Estimated annual family income \$ 600 - \$2500
c. Foreign-born None %; d. Negro Yes %; 1 %
(Nationality) (Yes or No)
e. Infiltration of None %; f. Relief families Few %
g. Population is increasing Slowly ; decreasing ; static

6. BUILDINGS: Small singles and
a. Type or types duplexes ; b. Type of construction Frame ;
c. Average age 12 - 15 years ; d. Repair Fair

7. HISTORY: SALE VALUES PREDOMINANT RENTAL VALUES
YEAR RANGE INATING % YEAR RANGE INATING %
1929 level \$1800 - \$6000 \$2500 100% \$20 - \$40 \$25 100%
1933 low \$1200 - \$4500 \$1800 70% \$15 - \$35 \$20 80%
current \$1800 - \$5000 \$2250 85% \$15 - \$40 \$25 100%
Peak sale values occurred in 1929 and were 100 % of the 1929 level.
Peak rental values occurred in 1929 and were 100 % of the 1929 level.

8. OCCUPANCY: a. Land 20 %; b. Dwelling units 98 %; c. Home owners 50 %

9. SALES DEMAND: a. Fair ; b. \$2250 singles ; c. Activity is Fair

10. RENTAL DEMAND: a. Good ; b. \$25 singles ; c. Activity is Good

11. NEW CONSTRUCTION: a. Types Small singles ; b. Amount last year Mediocre

12. AVAILABILITY OF MORTGAGE FUNDS: a. Home purchase Limited ; b. Home building Limited

13. TREND OF DESIRABILITY NEXT 10-15 YEARS Static

14. CLARIFYING REMARKS: Best portion along Chapel Hill Road and part of James Street

Example survey. (URichmond Mapping Inequality)

Energy Inequity is in part the result of a *hysteresis* effect rooted in historic housing discrimination.

Redlining was a "critical juncture" that separated otherwise similar housing stock.

- Test by examining differences in home energy services quality between redlined and observably similar households, measured as the
 - (1) *presence of sufficient heating technology* and
 - (2) *energy consumption responses to cold weather shocks*
 - Controlling for historic and current small neighborhood characteristics

Energy Inequity is in part the result of a *hysteresis* effect rooted in historic housing discrimination.

Redlining was a "critical juncture" that separated otherwise similar housing stock.

- Test by examining differences in home energy services quality between redlined and observably similar households, measured as the
 - (1) *presence of sufficient heating technology* and
 - (2) *energy consumption responses to cold weather shocks*
 - Controlling for historic and current small neighborhood characteristics

Not addressed here

- Lending discrimination debatably ended with CRA in 1977. Households able to migrate, re-sort. Why does Energy Inequity persist?
- Test for "stickiness" of neighborhood.
 - High non-market moving costs.
 - Neighborhood support, family proximity, etc.

Historic data → many assumptions

Prior literature

- Hoffman et al (2020) urban heat islands and redlined areas
- Nardone et al (2019) asthma and redlined areas
- Aaronson et al (2020) examined credit availability in redlined areas over 1930-1980 with RD-based analysis

Enlightening and incredibly inconvenient:

Fishback, La Voice, Shertzer, and Walsh (2020) on **endogeneity of redlining designation**.

- Used linked 1930 census address data and HOLC maps to show that redlined areas captured pre-existing economic and racial discontinuities in space.
- Border discontinuities not smooth in unobserveds. Even large moves in boundaries would still capture pre-existing segregations.
- Hillier (2003) no widespread proof that HOLC maps were distributed and used.

Empirical strategy

Acknowledging Fishback et al (2020), I control for selection on observables:

- Rent and income (historic and current)
- Presence of minorities in 193X
- Repair and quality of housing in 193X

Assume: conditional on observables that determined selection, Grade D (red) is as good as randomly assigned

- Unobserved neighborhood characteristics in 1930's not captured by observables are no longer relevant today.

Identification of effect of redlining using observably similar HOLC neighborhoods

- Multiple surveyors result in nearly arbitrary designation of Grade D (red) and Grade C (yellow) when "subversive minorities" were present.
- Many Grade C (yellow) areas had larger Black populations or worse home repair than nearby Grade D (red).

Empirical strategy

Acknowledging Fishback et al (2020), I control for selection on observables:

- Rent and income (historic and current)
- Presence of minorities in 193X
- Repair and quality of housing in 193X

Assume: conditional on observables that determined selection, Grade D (red) is as good as randomly assigned

- Unobserved neighborhood characteristics in 1930's not captured by observables are no longer relevant today.

Identification of effect of redlining using observably similar HOLC neighborhoods

- Multiple surveyors result in nearly arbitrary designation of Grade D (red) and Grade C (yellow) when "subversive minorities" were present.
- Many Grade C (yellow) areas had larger Black populations or worse home repair than nearby Grade D (red).
- Drawback: leaning on linear controls.
 - Solution: very flexible with linear controls.

(1) Presence of sufficient heating technology

- Ask "are there differences in heating sources between redlined and near-redlined homes, controlling for selection on observables?"
- Substandard heating technology = Coal or "None"
- Requires spatially explicit heating fuel use

(2) Energy consumption responses to cold weather shocks

- Ask "are there differences in energy consumption responses to weather shocks between redlined and near-redlined homes, controlling for selection on observables?"
- Data on monthly consumption and heating degree days for households in Grade D / Grade C areas
- Requires spatially explicit household electricity consumption with income

HOLC from URichmond "Mapping Inequality"

- 196 cities, 8,877 neighborhoods
- Survey data processed
 - Grade A-B-C-D
 - Repair class
 - Median Income 1936
 - Mean rent 1936
 - Presence of Blacks 1936

2018 ACS at block-group

- 44,357 BGs intersect HOLC
 - Heating fuel
 - Coal + "None" → substandard
 - Racial distribution
 - Median income 2018

HOLC from URichmond "Mapping Inequality"

- 196 cities, 8,877 neighborhoods
- Survey data processed
 - Grade A-B-C-D
 - Repair class
 - Median Income 1936
 - Mean rent 1936
 - Presence of Blacks 1936

Overlay BG with HOLC, keeping those BG that have >80% within one grade

- Take areal average when BG covers multiple HOLC neighborhoods of same grade
- 6,715 have most HOLC information

2018 ACS at block-group

- 44,357 BGs intersect HOLC
 - Heating fuel
 - Coal + "None" → substandard
 - Racial distribution
 - Median income 2018

All block-groups in Berkeley, CA



Measuring Hh response to temperature shocks

California RASS (Residential Appliance Saturation Survey)

- Confidential dataset with 24,216 homes surveyed in CA in 2009
 - Monthly consumption (from utility) for electricity, gas (if used)
 - Monthly HDD and CDD
 - Primary heating fuel
 - Income
 - Nighttime thermostat setpoint
 - Daytime thermostat setpoint
 - **Zip code**
- 138 households in 37 zip codes with >80% coverage for electric
- 1,018 households in 83 zip codes with >80% coverage for gas

Flexible fixed effect specification

$$\textit{PercentSubstandard}_b = \beta_0 + \sum_{g \in \{A, B, \}} \beta_g^{\textit{Grade}} + \beta \mathbf{x}_b + \Gamma_{c(b)} + \gamma_{c(b)} \mathbf{w}_b + \epsilon_b$$

- *PercentSubstandard* is share of 2018 homes with coal or no heating fuel in block-group b
- β_g is coefficient of interest
- \mathbf{x}_b is repair class, presence of Blacks in 1936
- $\Gamma_{c(b)}$ are county FEs for county c
- $\gamma_{c(b)}$ are county-specific slope shifters
- \mathbf{w}_b
 - Median income in 1936, 2018
 - Mean rent 1935
 - Presence of Blacks in 1936

Table 1: Share of Households with Substandard Fuel (Coal and None) by HOLC Grade

	Dependent Variable: Share of Households in Block Group with Substandard Heating			
	Model 1	Model 2	Model 3	Model 4
Grade D (Red)	0.00335* (0.00160)	0.00279+ (0.00155)	0.00285* (0.00143)	0.00416** (0.00135)
Grade B (Blue)	-0.00071 (0.00158)	-0.00050 (0.00161)	-0.00078 (0.00126)	0.00006 (0.00140)
Grade A (Green)	0.00005 (0.00280)	0.00045 (0.00289)	0.00042 (0.00239)	0.00180 (0.00322)
Predom. Black 2018			-0.00331** (0.00103)	-0.00223 (0.00144)
Predom. Asian 2018			-0.00408+ (0.00244)	-0.00172 (0.00185)
Predom. other race 2018			-0.00059 (0.00396)	0.00211 (0.00454)
Num.Obs.	6715	6715	3998	3998
R2 Adj.	0.126	0.121	0.070	0.070
FE: STCO	X	X	X	X
Control for home repair status 1935	X	X	X	X
County-specific slope on Med. Income 2018, 1936	X	X	X	X
County-specific slope on Mean Rent 1935	X	X	X	X
County-specific binary on Presence of Blacks 1935		X	X	X
Intx Predom. race 2018 and HOLC Grade				X

+ $p < 0.1$, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Robust SE clustered by FIPS county

Omitted Grade: C (Yellow)

Omitted Race: White

Response to temperature shocks

Home may have insufficient energy service quality if energy consumption responses to weather shocks are very large.

Response to temperature shocks

Home may have insufficient energy service quality if energy consumption responses to weather shocks are very large.

- Response is endogenous: both homes with efficient heating **and** inefficient homes who meet budget constraints with conservative thermostat settings have low consumption responses to weather shocks.
 - Observationally equivalent without observing thermostat setpoint.

$$\begin{aligned}
 consumption_{ht} = & \beta_0 + \beta_1 \quad_{ht} + \sum_{g \in \{A, B, \}} \beta_g \cdot \quad_{ht} \cdot 1(g = g(h)) \\
 & + \sum_{l=1}^3 \sum_{s=1}^5 \beta_{ls} \quad_{ht} \cdot 1(IG \quad TSET_h = s) \cdot 1(ClimateZone_h = l) + \\
 & + \beta_{inc} \cdot \quad_{ht} \cdot avgincome_h + \Gamma_h + \varepsilon_{ht}
 \end{aligned}$$

- $consumption_{ht}$ is energy (kWh, therms) consumption for household h month t
- $g(h)$ is HOLC Grade g for h
- \quad_{ht} is the heating-degree day for h in month t
- $IG \quad TSET_h$ is the thermostat setting for h
- $ClimateZone_h$ is the climate type for h
- $income_h$ is reported income for h
- Γ_h is household h fixed effect

Table 1: Regression of electricity consumption on heating degree days, interacted with HOLC grade and income, conditional on thermostat setpoint

	Dependent Variable: Energy consumption (kWh)		
	Model 1	Model 2	Model 3
hdd	3.398*** (0.806)	2.598*** (0.184)	
hdd x Grade D (Red)	3.737*** (0.573)	2.809*** (0.727)	3.057*** (0.673)
hdd x Avg rent 37-39	-0.050** (0.018)		-0.011 (0.013)
Num.Obs.	593	1150	593
R2 Adj.	0.804	0.826	0.802
FE: CZT24			X
FE: IDENT	X	X	X
Climate zone FE		X	X
Avg Inc x hdd	X	X	X
hdd x Thermostat setting		X	X
Thermostat setting x Climate Zone x hdd		X	X
Controls for 1937 incl. rent, presence of Blacks	X		X

+ $p < 0.1$, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Using only households with elec. as primary heating fuel

Omitted grade is "C"

Table 1: Regression of natural gas consumption on heating degree days, interacted with HOLC grade and income, conditional on thermostat setpoint

	Dependent Variable: Energy consumption (therms)		
	Model 1	Model 2	Model 3
hdd	0.311*** (0.089)	0.280* (0.111)	
hdd x Grade D (Red)	0.121 (0.244)	0.112 (0.245)	0.144 (0.247)
hdd x Avg rent 37-39	-0.001 (0.001)	-0.001 (0.002)	0.001 (0.001)
Num.Obs.	3623	3623	3623
R2 Adj.	0.564	0.563	0.556
FE: CZT24			X
FE: IDENT	X	X	X
Hdd x avg inc	X	X	X
hdd x thermostat setting		X	X
Hdd x thermostat setting x Climate Zone			X
Hdd x controls for 1937 incl. rent, presence of Blacks	X		X

+ p < 0.1, * p < 0.05, ** p < 0.01, *** p < 0.001

Using only households with natural gas as primary heating fuel

Omitted grade is "C"

Evidence of lingering differences in heating technology in/out of redlined areas

- Remains after controlling for observable differences in 193X
- Useful for targeting of energy efficiency programs

Evidence of larger consumption responses to cold weather shocks in redlined areas

- Conditional on 193X observables
- Conditional on thermostat setpoints

Further work

- Understanding selection into Grade D (red)
- "Stickiness" of redlined areas

Thanks!

jkirk@msu.edu