

Socioeconomic Expressions of Energy Poverty

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A dark blue diagonal gradient bar that starts from the bottom left and extends towards the top right, covering the lower half of the slide.

Background



Do you turn off appliances when not in use?

Do you wear layers indoors instead of using heat?

Do you use a cold rag when it's hot out instead of turning on the AC?



Research Questions

- How do poverty status in 2020 and other socioeconomic factors impact people's ability to manage their energy consumption and, consequently, their health?
- Does energy poverty status relate to demographics such as race and income status?
- Do respondents require medical attention due to extreme heat or cold in their homes?

Literature

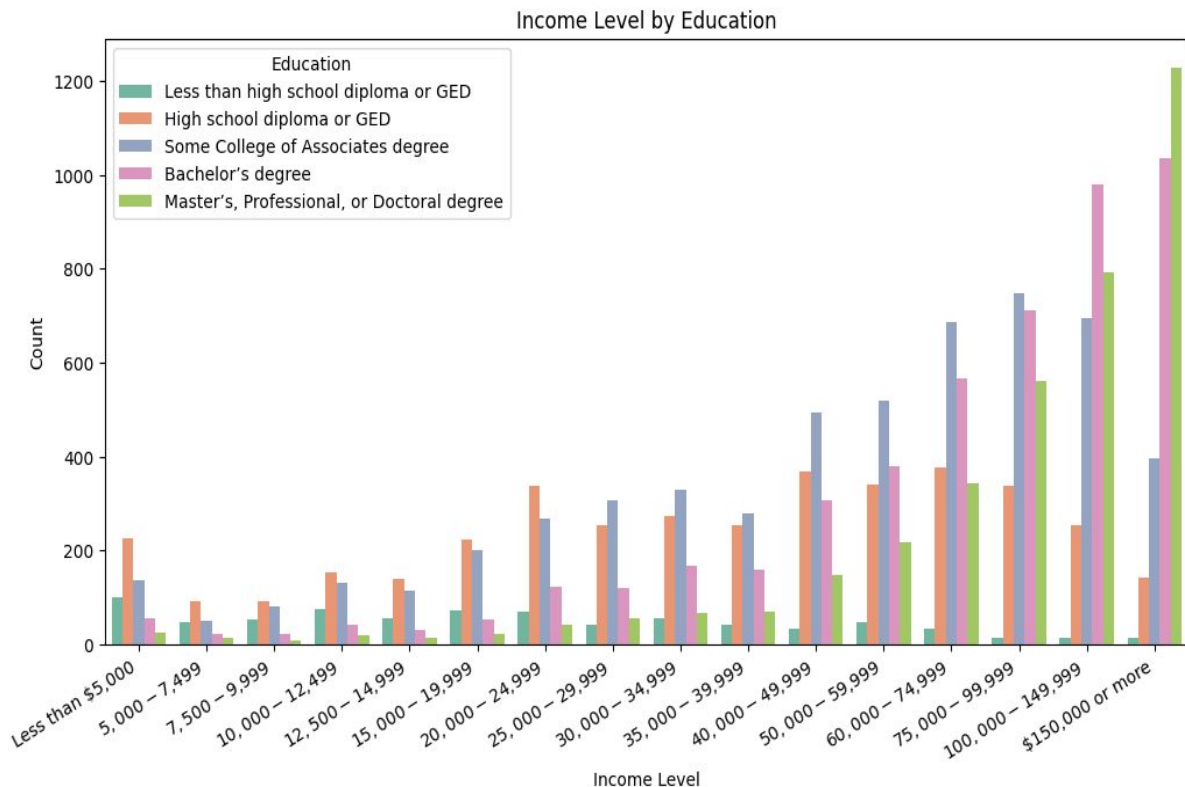
- Li et al. (2021):
 - Studied the relationship between energy poverty and energy efficiency; discovered that energy poverty reduces GDP
 - Over time, results in sharp declines in a nation's social welfare and low-income families are typically the ones who are directly affected
- Halkos and Gkampoura (2021):
 - Energy poverty was linked to energy prices, unemployment, and the number of persons at risk of poverty
 - GDP is inversely associated to energy poverty, which supports the findings of Li et al. (2021)
- Recalde et al. (2019):
 - Introduced Structural Energy Poverty Vulnerability (SEPV) index
 - Countries with lower SEPV scores experienced significantly higher rates of energy poverty and increased risk of excess winter mortality
- Pan, Biru, and Lettu (2021):
 - Energy poverty has a harmful impact on public health
 - Higher living standards lessen the effects

Data Overview



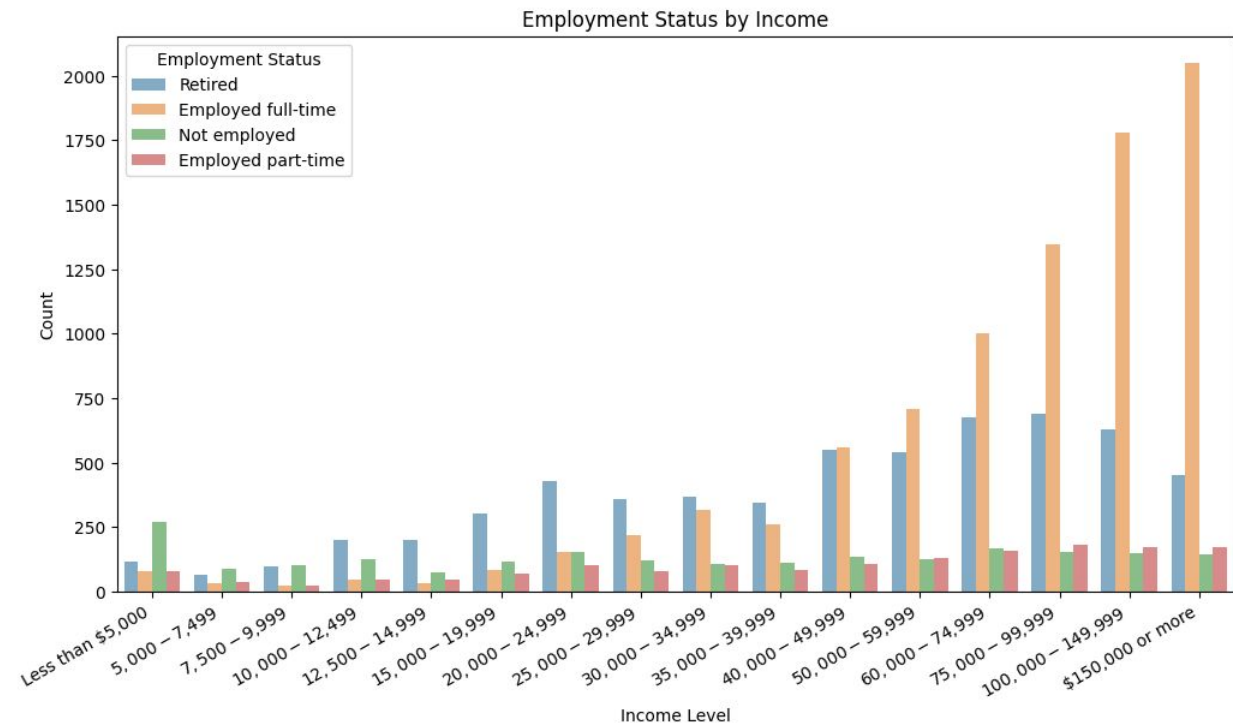
- US Energy Information Administration's (EIA)
- Residential Energy Consumption Survey (RECS)
 - Survey conducted every 5 years
 - 700 variables
 - ~18,500 households surveyed
 - 2020 Survey

Exploratory Data Analysis



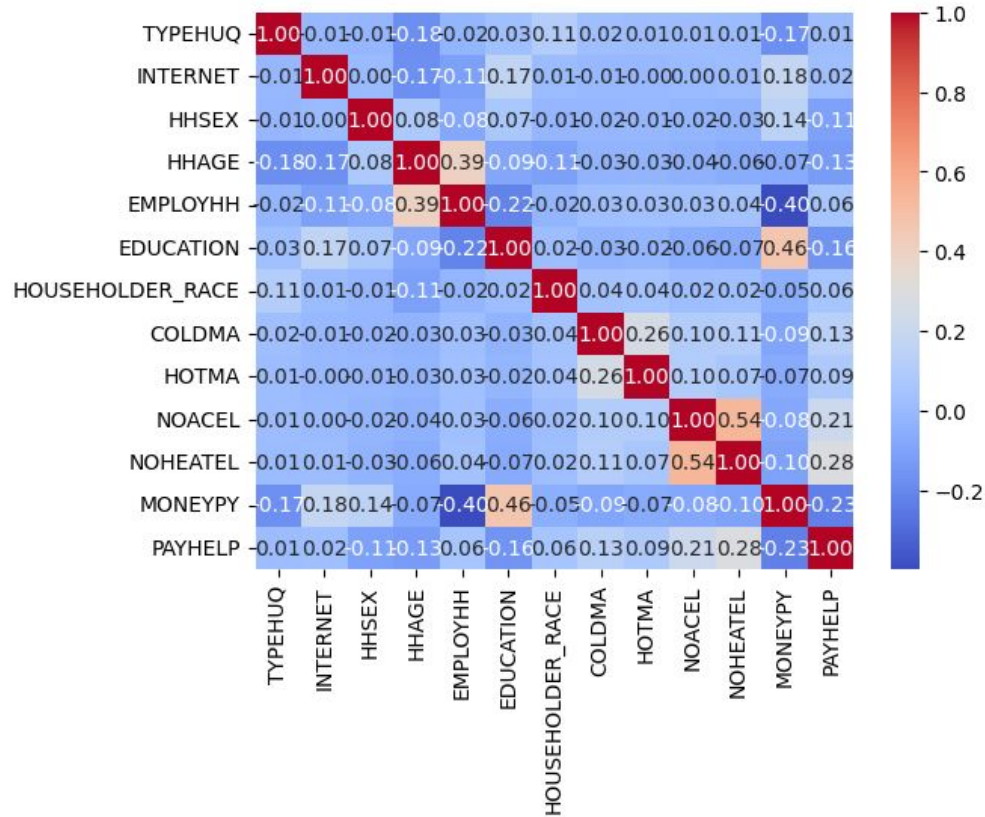
- The plot shows the correlation between education and income.
 - a. Higher incomes correlated with advanced degrees.
 - b. “\$100,000 - \$149,999”; “\$150,000 or more,” are strongly correlated with “Bachelor’s,” “Master’s,” or “Doctoral” degrees.

Exploratory Data Analysis: cont.



- The plot between employment status and income.
 - a. Higher incomes correlated with full-time employment.
 - b. Upper incomes are strongly correlated with “Employed Full Time” and “Retired” statuses.

Correlation Matrix



Random Forest

$\text{Payhelp} = \alpha + \text{'MONEYPY'} +$
 $\text{'HOUSEHOLDER_RACE'} + \text{'EDUCATION'} +$
 $\text{'TYPEHUQ'} + \text{'INTERNET'} + \text{'HHSEX'} +$
 $\text{'HHAGE'} + \text{'EMPLOYHH'} + \text{'COLDMA'} +$
 $\text{'HOTMA'} +, \text{'NOACEL'} + \text{'NOHEATEL'} + \epsilon$

Accuracy 88.78%

[[**3258**, FP: 95, FP: 14],

[FN: 204, **23**, FP: 13],

[FN: 71, FN: 18, **4**]]

Logistic Regression

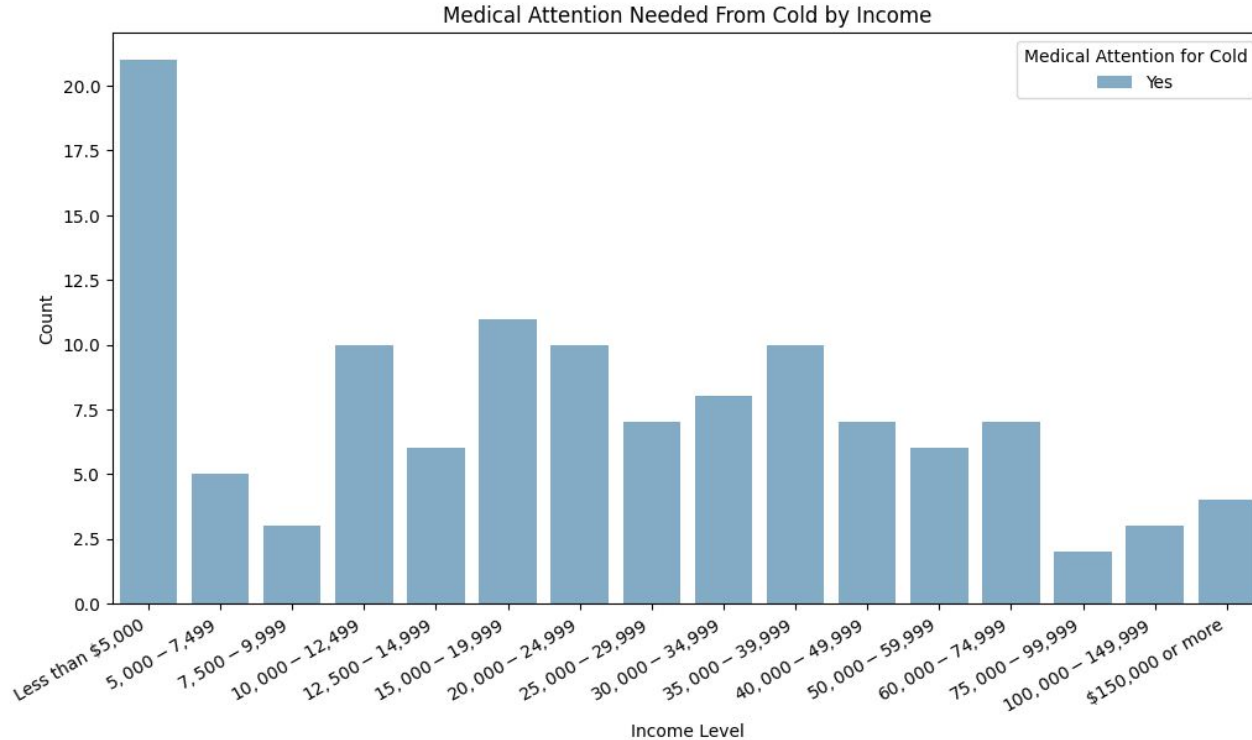
Table 6: Results: Generalized linear model

Model:	GLM	AIC:	1672.1472
Link Function:	Logit	BIC:	-10272.9646
Dependent Variable:	['PAYHELP[No]', 'PAYHELP[Yes]']		Log-Likelihood: -817.07
Date:	2023-12-07 21:18		LL-Null: -910.23
No. Observations:	1629		Deviance: 1634.1
Df Model:	18		Pearson chi2: 1.61e+03
Df Residuals:	1610		Scale: 1.0000
Method:	IRLS		

		Coef.	Std.Err.	z	P> z	[0.025	0.975]
Not Large Enough 168% - 275% for "Yes"	Intercept	-0.2442	0.3090	-0.7904	0.4293	-0.8498	0.3614
	(MONEYPY)[Less than \$5,000]	0.2766	0.2766	1.0001	0.3173	-0.2655	0.8186
	(MONEYPY)[\$5,000 - \$7,499]	0.2048	0.3678	0.5569	0.5776	-0.5161	0.9258
	(MONEYPY)[\$7,500 - \$9,999]	-0.5109	0.3700	-1.3808	0.1673	-1.2361	0.2143
	(MONEYPY)[\$12,500 - \$14,999]	0.3755	0.3321	1.1304	0.2583	-0.2755	1.0264
	(MONEYPY)[\$15,000 - \$19,999]	0.5219	0.3166	1.6482	0.0993	-0.0987	1.1425
	(MONEYPY)[\$20,000 - \$24,999]	0.8487	0.3014	2.8155	0.0049	0.2579	1.4394
	(MONEYPY)[\$25,000 - \$29,999]	1.0136	0.3057	3.3156	0.0009	0.4144	1.6128
	(MONEYPY)[\$30,000 - \$34,999]	0.7587	0.3039	2.4967	0.0125	0.1631	1.3543
	(MONEYPY)[\$35,000 - \$39,999]	1.1165	0.3189	3.5017	0.0005	0.4916	1.7415
305% - 1869% for "No"	(MONEYPY)[\$40,000 - \$49,999]	1.3453	0.3013	4.4650	0.0000	0.7548	1.9359
	(MONEYPY)[\$50,000 - \$59,999]	1.6126	0.3134	5.1458	0.0000	0.9984	2.2268
	(MONEYPY)[\$60,000 - \$74,999]	1.9722	0.3358	5.8731	0.0000	1.3140	2.6303
	(MONEYPY)[\$75,000 - \$99,999]	2.5634	0.4249	6.0332	0.0000	1.7306	3.3961
	(MONEYPY)[\$100,000 - \$149,999]	2.9280	0.5558	5.2682	0.0000	1.8387	4.0174
	(MONEYPY)[\$150,000 or more]	2.0196	0.5694	3.5471	0.0004	0.9036	3.1355
	(HOTMA)[Yes]	0.2156	0.4007	0.5379	0.5906	-0.5698	1.0009
	(COLDMA)[Yes]	-0.8172	0.2914	-2.8048	0.0050	-1.3882	-0.2461
	HHAGE	0.0082	0.0042	1.9796	0.0477	0.0001	0.0164

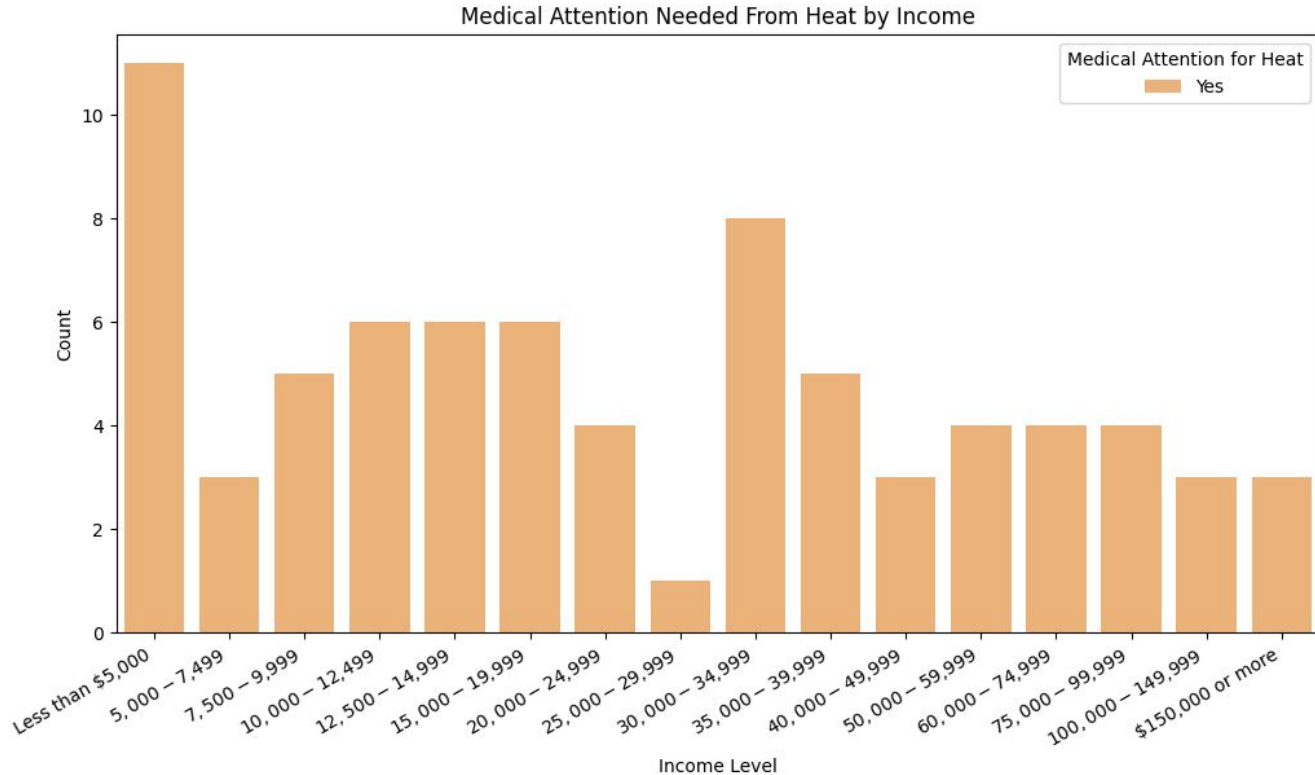
44% →

Chi-Square Tests



Chi-squared statistic: 111.83 Degrees of freedom: 15 P-value: 7.16e-17

Chi-Square Tests



Chi-squared statistic: 185.85; Degrees of freedom: 15; P-value: 1.56e-31

Other Tests We Tried (but did not use)

Linear Regression:

$$\text{'COLDMA'} = \alpha + \text{'MONEYPY'} + \text{'HOUSEHOLDER_RACE'} + \text{'EDUCATION'} + \epsilon$$

Multi-Linear Regression:

$$\text{'COLDMA'} = \alpha + \text{'MONEYPY'} + \text{'HOUSEHOLDER_RACE'} + \text{'EDUCATION'} + \text{'NOACEL'} + \text{'NOHEATEL'} + \text{'HHAGE'} + \text{'PAYHELP'} + \epsilon$$

Logistical:

$$\text{'COLDMA'} = \alpha + \text{'MONEYPY'} + \text{'HOUSEHOLDER_RACE'} + \text{'EDUCATION'} + \text{'NOACEL'} + \epsilon$$

$$\text{'HOTMA'} = \alpha + \text{'EDUCATION'} + \text{'HOUSEHOLDER_RACE'} + \text{'NOACEL'} + \text{'MONEYPY'} + \epsilon$$

SVM:

$$\text{'PAYHELP'} = \alpha + \text{'MONEYPY'} + \text{'EDUCATION'} + \text{'EMPLOYHH'} + \text{'TYPEHUQ'} + \text{'NOACEL'} + \text{'NOHEATEL'} + \epsilon$$

Conclusion

- Lower incomes have higher rates of energy poverty.
- COLDMA has a stronger influence overall than HOTMA.
- This could be due to lower costs and easier access to cooling measures than the various fuels needed for heating.
- Our findings reflect conclusions from the existing literature.

Limitations and Future Steps

- Time series data (panel data)
- Examine COVID-19 effects in the RECS survey
- State or county level could be different

Thank You



References

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