

ReThinking Emergency Rental Allocation

Appendix C. Econometric models descriptors

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Being at risk of eviction is a result of several factors. Recognizing the multicasual nature of this social phenomenon, the *ReThinkers* team designed a quantitative strategy sensitive to both personal-level factors and county-level factors.

Datasets

- Household Pulse Survey (HSP)
- Census data
- State eviction laws data sets
- Treasury Emergency Rental Assistance (ERA) Dashboard

Software

For estimation of econometric modeling we used the following R-Packages:

- A grammar of data manipulation (`dplyr` package[@wickham2015dplyr])
- Elegant graphics for data analysis (`ggplot2` package[@wickham2011ggplot2])
- Analysis and Presentation of Social Scientific Data (`jtools` package[@long2017package])
- Econometric modeling (`stats` package[@R-core])
- GAM and Generalized Ridge Regression for R (`mgcv` package[@wood2012mgcv])

`dplyr`

A grammar of data manipulation, `dplyr` package, was built by Hadley Wickham. It “provides a consistent set of verbs that help you solve the most common data manipulation challenges”.[@wickham2015dplyr] We used `dplyr` for the initial data processing in order to have clean data suited for the modeling and visualization processes.

ggplot2

All the visualizations in this report were built using the **ggplot2** package, created by Hadley Wickham, which is “a system for declaratively creating graphics, based on The Grammar of Graphics” by Leland Wilkinson[@wilkinson2012grammar].

epitools

We also took advantage of the **epitools** package, a set of “basic tools for applied epidemiology”.[@aragon2012epitools] We used the package to estimate the observed variation of homicide data.

lme4

All the models from the first section were built using the **lme4** package, created by Ben Bolker. It is designed to fit Linear, Generalized Linear, And Nonlinear Mixed models, providing different functions such as **lmer()**, **glmer()**, and **nlmer()**.[@bates2007lme4] Since we are modeling counts and rates (assuming a Poisson distribution for the dependent variable), we used the **glmer()** function, fitted for Generalized Linear Mixed Models (GLMM).

jtools

Jacob A. Long wrote the **jtools** package as a set of tools “for the purpose of more efficiently understanding and sharing the results of regression analyses”.[@long2017package] For our analysis, we used the **summ()** function to present the results of the **glmer()** models. The **summ()** command “prints output for a regression model in a fashion similar to **summary**, but formatted differently with more options”, which include including data description, goodness of fit measurements, as well as fixed and random effects coefficients.[@long2017package]

merTools

The **merTools** package, Tools for Analyzing Mixed Effect Regression Models, “provides methods for extracting results from mixed-effect model objects fit with the **lme4** package” and “allows construction of prediction intervals efficiently from large scales linear and generalized linear mixed-effects models”.[@knowles2016mertools] From this package, we used the **predictInterval** command, which provides the confidence intervals for the random effects coefficients using bootstrapping techniques.

mgcv

Finally, we use the **mgcv** package, Mixed GAM Computation Vehicle with Automatic Smoothness Estimation, by Simon Wood to perform final modeling with the smoothness transformations.[@wood2012mgcv] In particular, we used the **gamm()** function for modeling Generalized Additive Mixed Models with the smoothness (achieved through the use of splines) applied to the *time* variable.

Probit model for personal-level factors

The following snippet code shows the probit model specification.

```
# Model
m1 <- glm(
  eviction_likelihood ~ age + female + white + black + asian +
    hispanic + non_het + trans + single + kids + college +
    unemployment_month + military + disability_phys + disability_mental +
    rent_assistance + rent_increase + rent_delay_months + (single*kids),
  family = binomial(link = "probit"),
  weights = df_clean$pweight,
  data = df_clean,
  # na.rm = T
)
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

Here's the table of coefficient.

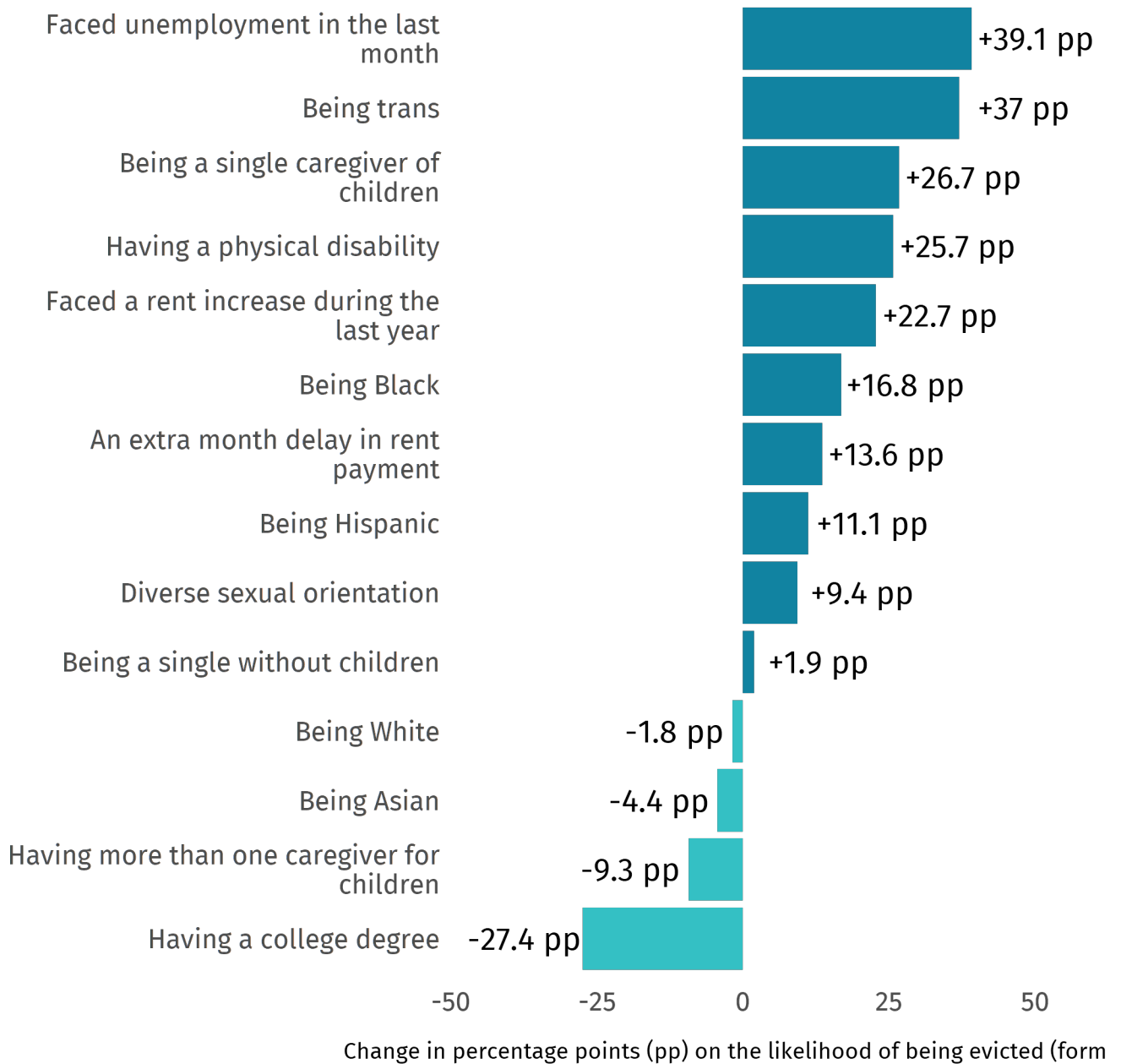
```
##
## Call:
## glm(formula = eviction_likelihood ~ age + female + white + black +
##      asian + hispanic + non_het + trans + single + kids + college +
##      unemployment_month + military + disability_phys + disability_mental +
##      rent_assistance + rent_increase + rent_delay_months + (single *
##      kids), family = binomial(link = "probit"), data = df_clean,
##      weights = df_clean$pweight)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -331.53  -35.84   -6.09   35.47  290.99
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -5.9060104  4.2964095  -1.375  0.1692
## age           -0.0001489  0.0000469  -3.175  0.0015 **
## female        -0.1828730  0.0011886 -153.853 < 2e-16 ***
## white         -0.0434406  0.0021189 -20.502 < 2e-16 ***
## black          0.1434287  0.0021817  65.741 < 2e-16 ***
## asian         -0.0194940  0.0030307  -6.432 1.26e-10 ***
## hispanic       0.1314997  0.0013926  94.428 < 2e-16 ***
## non_het        0.1084790  0.0015415  70.374 < 2e-16 ***
## trans          0.3320077  0.0055782  59.519 < 2e-16 ***
## single         0.0037670  0.0016866   2.233  0.0255 *
## kids          -0.0802096  0.0019753 -40.607 < 2e-16 ***
## college       -0.2647078  0.0013700 -193.213 < 2e-16 ***
## unemployment_month 0.3799888  0.0011967 317.528 < 2e-16 ***
## military       5.0417711  4.2964074   1.173  0.2406
## disability_phys 0.1881264  0.0012715 147.958 < 2e-16 ***
## disability_mental 0.1146980  0.0012116  94.669 < 2e-16 ***
## rent_assistance 0.1407315  0.0016351  86.071 < 2e-16 ***
## rent_increase  0.2144263  0.0011359 188.775 < 2e-16 ***
## rent_delay_months 0.1365571  0.0002486 549.245 < 2e-16 ***
## single:kids     0.2613878  0.0024008 108.878 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 4521035  on 860  degrees of freedom
## Residual deviance: 3430427  on 841  degrees of freedom
##      (50076 observations deleted due to missingness)
## AIC: 3434510
##
## Number of Fisher Scoring iterations: 6
```

And a more intuitive way of understanding significant results.

How does the likelihood of being evicted change?

For significant eviction-risk personal factors



Source: U.S. Census Bureau Household Pulse Survey, Week 49

Note: Total Population 18 Years and Older

Data processed by ReThinkers team for the Logistics challenge at the MIT Policy Hackathon 2022