# Further Examination of the Relationship Between Race and Gender in Mortgage Lending within the State of

Oklahoma. \*

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#### **Abstract**

The project answers a question about lending in Oklahoma; do mortgage rates differ among race or gender? The difference in difference analysis looks at data before and after the COVID 19 pandemic. Using linear regression the data-set is analyzed to reveal differences. (notfinished).

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#### 1 Introduction

The purpose of this report is to examine the relationship between the amount of a mortgage and the demographic characteristics of the borrower before and after the COVID 19 pandemic. The study uses data collected by the Federal Housing Finance Agency from a random sample of loan-level mortgage acquisitions acquired in 2019 and 2021. The analysis employs multiple linear regression models to estimate the impact of several independent variables, including race, gender, age, income, credit score, and loan-to-value ratio, on the dependent variable, the amount of the mortgage note/percentage rate.

## 2 Literature Review

Previous work by [1] shows that educational decisions are an important determinant of later-life earnings. This point is driven further in follow-up work by [1] and [1].

#### 3 Data

The primary data source for this research is the FHL Bank Public Use Database. Table 1 contains summary statistics.

## 4 Empirical Methods

While my approach explores a number of different approaches, the primary empirical model can be depicted in the following equation:(I will code math properly for final report!) [1] LTV log-

noteamt + noteratepercent + LTV + bo1race\*year + bo1gender\*year + bo1age + debtexpenseratio

[2] lognoteamt noteratepercent + LTV + bo1race\*year + bo1gender\*year + bo1age + debtexpenseratio + hsexpenseratio [3] noteratepercent debtexpenseratio + lognoteamt\*year + LTV + bo1race\*year + bo1gender\*year + bo1age

$$Y_{it} = \alpha_0 + \alpha_1 Z_{it} + \alpha_2 X_{it} + \varepsilon, \tag{1}$$

where  $Y_{it}$  is a continuous outcome variable for unit i in year t, and  $Z_{it}$  are characteristics about the firm at which i is working, while  $X_{it}$  are characteristics about i. The parameter of interest is  $\alpha_1$ .

#### **5** Research Findings

The main results are reported Figure 1, 2, 3

## 6 Conclusion

Here I will add my conclusion on the regression analysis. I did not get this far since I've been doing the data search, wrangling and now the difference in difference analysis.

## References

[1] Arpit Gupta, Christopher Hansman, and Pierre Mabille. Financial constraints and the racial housing gap. 2022.

	Unique (#)	Missing (%)	Mean	SD	Min	Median	Max	
totmonthlyincome	1090	0	9249.2	8286.5	1138.0	7260.0	110741.0	<b>L</b>
LTV	82	0	79.3	16.2	7.0	80.0	104.0	
bo1age	68	0	46.6	14.7	20.0	45.0	90.0	
noteratepercent	26	0	4.0	0.5	2.9	3.9	5.8	_علاد_
noteamt	783	0	194408.9	102657.3	26800.0	172000.0	484350.0	
hsexpenseratio	950	0	18.9	7.9	0.0	18.2	78.0	<b>.</b>
debtexpenseratio	972	0	31.0	9.0	1.1	31.4	78.0	
lognoteamt	783	0	12.0	0.5	10.2	12.1	13.1	
logtotmonthincome	1090	0	8.9	0.6	7.0	8.9	11.6	

Figure 1: 2019 data

# **Figures and Tables**

	Unique (#)	Missing (%)	Mean	SD	Min	Median	Max	
totmonthlyincome	1002	0	9838.6	8693.7	1342.0	8016.5	141878.0	<b>L</b>
LTV	80	0	75.6	15.0	16.0	80.0	104.0	
bo1age	65	0	47.1	13.4	20.0	45.0	86.0	_
noteratepercent	15	0	2.8	0.3	2.0	2.9	3.6	
noteamt	647	0	212522.4	115728.0	29000.0	182760.5	548250.0	
hsexpenseratio	855	0	18.0	7.9	0.0	16.9	50.2	_
debtexpenseratio	882	0	30.6	9.2	5.6	31.2	64.5	
lognoteamt	647	0	12.1	0.6	10.3	12.1	13.2	
logtotmonthincome	1002	0	9.0	0.6	7.2	9.0	11.9	

Figure 2: 2021 data

	(1)	(2)	(3)
(Intercept)	-10.623	11.414***	4.907***
	(7.541)	(0.140)	(0.267)
lognoteamt	6.636***		-0.113***
	(0.554)		(0.022)
noteratepercent	4.855***	-0.078**	
	(0.740)	(0.028)	
bo1race2	-7.545*	0.314**	-0.124
	(3.058)	(0.114)	(0.087)
bo1race3	-1.252	0.048	-0.049
	(3.206)	(0.119)	(0.091)
bo1race4	10.529	0.037	0.682+

Figure 3: regression results

```
Residuals:
        1Q Median
  Min
                      3Q
                            Max
-66.639 -7.178 0.954 9.303 37.741
Coefficients: (1 not defined because of singularities)
           Estimate Std. Error t value Pr(>|t|)
(Intercept)
              -10.62301
                         7.54128 -1.409 0.15908
                6.63556 0.55365 11.985 < 2e-16 ***
lognoteamt
                           0.73952 6.565 6.47e-11 ***
noteratepercent
                  4.85497
               -7.54544 3.05848 -2.467 0.01370 *
bo1race2
bo1race3
               -1.25218
                          3.20649 -0.391 0.69620
               10.52857 13.97287 0.754 0.45123
bo1race4
               -4.76355 1.43934 -3.310 0.00095 ***
bo1race5
year2021
               -0.06996
                         2.25285 -0.031 0.97523
bo1gender2
                 1.77288    0.88610    2.001    0.04554 *
               -0.31515
                         0.02127 -14.817 < 2e-16 ***
bo1age
debtexpenseratio
                   0.29758
                            0.03314 8.979 < 2e-16 ***
bo1race2:year2021
                    9.80301
                             4.65489 2.106 0.03532 *
bo1race3:year2021 0.88545
                             4.48463 0.197 0.84350
bo1race4:year2021
                       NA
                              NA
                                    NA
                                           NA
bo1race5:year2021
                    1.98127
                             2.10128 0.943 0.34584
                               1.28266 -0.545 0.58591
year2021:bo1gender2 -0.69886
Signif. codes: 0 "*** 0.001 "** 0.01 "* 0.05 ". 0.1 " 1
Residual standard error: 13.89 on 2194 degrees of freedom
```

Multiple R-squared: 0.226, Adjusted R-squared: 0.221 F-statistic: 45.75 on 14 and 2194 DF, p-value: < 2.2e-16

Figure 4: LTV regression results

```
Call:
```

```
Im(formula = lognoteamt ~ noteratepercent + LTV + bo1race * year +
bo1gender * year + bo1age + debtexpenseratio + hsexpenseratio,
data = df)
```

#### Residuals:

```
Min 1Q Median 3Q Max -1.65680 -0.35408 -0.00591 0.37239 1.65926
```

```
Coefficients: (1 not defined because of singularities)
            Estimate Std. Error t value Pr(>|t|)
              11.4138188 0.1399465 81.558 < 2e-16 ***
(Intercept)
                 -0.0784339 0.0277452 -2.827 0.00474 **
noteratepercent
LTV
             0.0093461 0.0007699 12.139 < 2e-16 ***
                0.3137623 0.1138377 2.756 0.00590 **
bo1race2
bo1race3
                0.0484186 0.1193184 0.406 0.68493
                0.0370061 0.5200609 0.071 0.94328
bo1race4
                0.1467065 0.0536095 2.737 0.00626 **
bo1race5
               0.0312175 0.0838709 0.372 0.70977
vear2021
                -0.2154022 0.0327097 -6.585 5.67e-11 ***
bo1gender2
              -0.0008992 0.0008304 -1.083 0.27896
bo1age
                  0.0010746 0.0014860 0.723 0.46966
debtexpenseratio
                 0.0071392 0.0016733 4.267 2.07e-05 ***
hsexpenseratio
bo1race2:year2021 -0.2196751 0.1733272 -1.267 0.20515
bo1race3:year2021 -0.0122286 0.1669414 -0.073 0.94161
bo1race4:year2021
                        NA
                               NA
                                      NA
                                            NA
bo1race5:year2021 -0.0161812 0.0782071 -0.207 0.83611
year2021:bo1gender2 0.0717689 0.0477083 1.504 0.13264
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.5167 on 2193 degrees of freedom Multiple R-squared: 0.1249, Adjusted R-squared: 0.1189 F-statistic: 20.87 on 15 and 2193 DF, p-value: < 2.2e-16

Figure 5: lognoteamount regression results

```
Call:
Im(formula = noteratepercent ~ debtexpenseratio + lognoteamt *
  year + LTV + bo1race * year + bo1gender * year + bo1age,
  data = df
Residuals:
         1Q Median
                        3Q
  Min
                              Max
-1.10116 -0.26905 -0.02616 0.25431 1.84979
Coefficients: (1 not defined because of singularities)
            Estimate Std. Error t value Pr(>|t|)
              4.9073518 0.2671989 18.366 < 2e-16 ***
(Intercept)
debtexpenseratio
                  0.0035498 0.0009580 3.705 0.000216 ***
               -0.1127740 0.0222035 -5.079 4.11e-07 ***
lognoteamt
              -2.7785186 0.3797667 -7.316 3.56e-13 ***
year2021
LTV
             0.0040004 0.0006021 6.644 3.85e-11 ***
bo1race2
               -0.1236231 0.0872952 -1.416 0.156874
              -0.0494412 0.0913110 -0.541 0.588246
bo1race3
               0.6816855 0.3977135 1.714 0.086668 .
bo1race4
bo1race5
               0.0287699 0.0411125 0.700 0.484136
                -0.0466421 0.0254411 -1.833 0.066888 .
bo1gender2
              -0.0004380 0.0006352 -0.690 0.490559
bo1age
lognoteamt:year2021 0.1332211 0.0311411 4.278 1.97e-05 ***
vear2021:bo1race2 0.0471251 0.1328974 0.355 0.722925
vear2021:bo1race4
                       NA
                              NA
                                    NA
                                           NA
year2021:bo1race5  0.0045826  0.0599208  0.076  0.939047
year2021:bo1gender2 0.1047463 0.0368786 2.840 0.004549 **
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.3954 on 2193 degrees of freedom Multiple R-squared: 0.6846, Adjusted R-squared: 0.6825 F-statistic: 317.4 on 15 and 2193 DF, p-value: < 2.2e-16

Figure 6: note rate percentage regression results