# Eviction Court Cases

# Douglas & Lancaster Counties, Nebraska

# May 2, 2023

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## 1 Documentation, Code and Data Repository

This is a formal documentation of the efforts to collect eviction court cases in Lancaster and Douglas Counties in the state of Nebraska. The code and data repositories are stored in the links below

- 1. Code Repository
- 2. Data Repository

## 2 Case Information

The scraped court cases from nebraska.gov/justice/case.cgi are stored in the Data Repository as clean\_all\_cases.csv, where the data is cleaned and extracted. Specifically, we were interested in the

- 1. Case Summary
- 2. Parties/Attornies to the Case
- 3. Judgement Information
- 4. Register of Actions

From the given fields, we were able to extract relevant information that include but not limited to

- 1. Case Name
- 2. Plaintiff
- 3. Defendant
- 4. Judge
- 5. Classification
- 6. Attorneys
- 7. Defendant's Address
- 8. Decision

Note that the information is extracted via regular expression, and hence is not guaranteed to be accurate for all cases. We would only further discuss some aspects of the case information where certain choices were made due to the nature of the data.

### 2.1 Classification

We were interested in the Classification of the court cases due to our efforts in scrape all court cases in both Lancaster and Douglas Counties as described in Section 5. However, since we are only interested in eviction cases, we need some methods of filtering the scraped cases.

Based on comprehensive list of eviction cases compiled by Professor Sullivan, which is simplified to a csv that can be found in here, we get the following frequency table for classifications of court cases in Table 1.

Classification	Frequency
Contract-Unspecified	1
Real Property-Forcible Entry/Det-Damages	2,652
Real Property-Forcible Entry/Detainer	23,391
Real Property-Landlord/Tenant	34,055
Real Property-Landlord/Tenant-Damages	808
Real Property-Unspecified	2

Table 1: Classification of Cases

While I am not entirely certain that the **Contract-Unspecified** is an error or otherwise, we are going to only include cases that specificy that it is **Real-Property** related. We also do not distinguish the type of Real-Property cases.

## 2.2 Writ and Judgement of Restitution

We wish to identify cases where writ of restitution was issued as an indicator if the eviction process was carried out. The information is obtained from the **Register of Actions** tab in the court case.

However, we were informed by Professor Sullivan that the **Register of Actions** tab is manually entered. There were cases were *restitution* were misspelled, and hence the initial strategy of performing a regex search on

#### writ of restitution

yielded quite a few false negatives. Hence, the current (as of May 2, 2023) approach is to search for the term

writ

though this might introduce false positives as a result.

Nevertheless, there is a corresponding Judgement of Restitution under the **Judgement Information** tab, which is automated and *perhaps* more consistent. However, a point to note is that the issuance of writ of restitution under **Register of Actions** and Judgement of Restitution under **Judgement Information** can be a different. Table 2 shows the difference in terms of their joint distributions.

	Judgement	NO Judgement	
Writ of Restitution NO Writ of Restitution	$0.609 \\ 0.113$	$0.004 \\ 0.275$	.613 .387
	0.722	0.278	

Table 2: Joint Probabilities of Judgemeent and Writ of Restitution

### 2.3 Continuance

We are interested if the court case is continued, which prolongs the case which theoretically, would allow the defendant to adjust in the event of an eviction. We obtain the data from the **Register of Actions** tab, and some of the observed words used are

Hence, we consider the case is continued if the words

#### continue or continuance

appears in the **Register of Actions** tab. However, there might be words that reflect is continuance but is not captured by the algorithm.

Is Continued?	Frequency	Proportion	
FALSE	49,032	0.805	
TRUE	11,877	0.195	

Table 3: Cases with an Issue of Continuance

### 2.4 Court Attendance

TODO

### 2.5 Limited Representation Attorney

TODO

## 3 Defendant Demographics

We are interested in the demographics of the defendant of the eviction case. Nevertheless, we are not given the demographic information from the court cases. However, using the R-package predictrace, we predict demographic information by looking at the named defendant in the case (there can be multiple defendants for a given case).

Gender	Frequency		
female female, male	30262		
male	26672		

Table 4: Predicted Gender Based on First Name

The algorithm is trained based on Census Data, where ethnicity and race is reported whereas gender is trained on Social Security Administration data.

	Last Name				
First Name	native	asian	black	hispanic	white
asian	0	93	6	16	62
asian, hispanic	0	0	0	0	1
asian, white	0	0	1	2	3
black	0	10	101	15	610
black, white	0	0	1	0	10
hispanic	3	18	58	1,546	510
hispanic, white	0	0	0	3	0
white	53	218	1,920	2,051	$30,\!412$

Table 5: Predicted Race Distribution Based on First Name and Last Name

Table 5 shows the frequency table for each of the race. Note that last names and first names are separately estimated. Also note that the categories are different.

There is an alternative specification of the algorithm, where the probability is provided instead. However, due to the differences in categories, I do not want to assume how to recategorize the demographic information.

## 4 Geocoding

For most of the cases scraped, the defendant's address is stated in the **Parties/Attorneys** to the Case tab. Using this stated address, we use the Census Geocoder to map the given address and match it to the Census Tracts based on the 2020 Census.

Exactness of Match	Frequency	Proportion
Exact	47,279	0.85
Not Exact	8,600	0.15

Table 6: Exactness of Match for Addresses

We were able to match 55,879 addresses out of 60,909 cases, which gives us a matching rate of 91.74%. The frequency and proportion of exactness for those which are matched can be seen in Table 6. I would not be concerned with the exactness since the differences are usually due to

- 1. Short forms street and st, avenue and ave, N and North etc.
- 2. Apartment Numbers
- 3. Difference in ZIP Code by a number

# 5 Brute Force Scraping

TODO