Documentation of Work + Insights

EDA/Data Cleaning File

Simple Data Structuring/Exploring

Process

- 1. Data started off as 4,346 rows with 10 columns.
 - Removed Null entries and empty columns → 3,190 rows and 9 columns.
- 2. Converted 'date' column into 2 new columns: ['new_date', 'new_time'].
 - Data type is **datetime**.
 - Removed 'date' column.
 - Data shape is 3,190 rows with 10 columns.

Visualizations

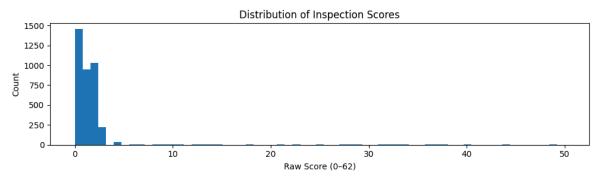


Figure I: Distribution of initial score range (0-62) of all 3,190 entries.

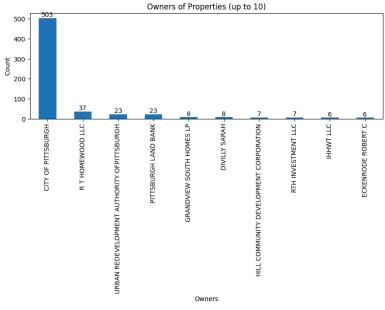
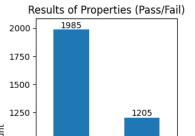


Figure II: Distribution of owners of the properties.



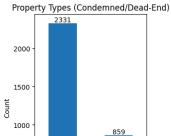


Figure III: Distribution of Pass/Fail Results.

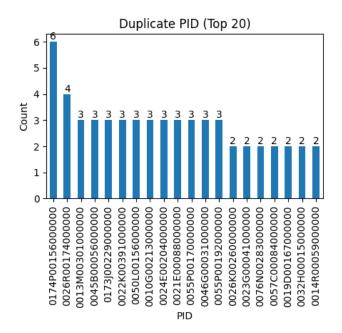
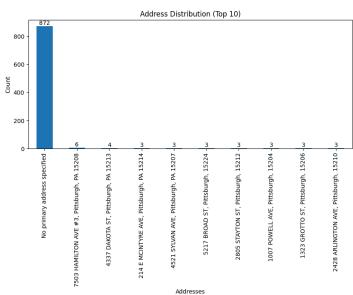


Figure VI: Distribution of addresses in the dataset.

Figure IV: Distribution of Condemned / Dead-End Properties.

Figure V: Distribution of PID's of all Properties (showing duplicate issue).



Insights

Figure I

• Distribution of the initial scores range mostly around 0-5.

Figure II

503 properties (~15.8% of properties) are owned by the City of Pittsburgh

Figure III

- ~62.2% of properties failed the latest inspections (1985 properties)
- ~37.8% of properties passed the latest inspections (1205 properties)

Figure IV

- ~73.1% of properties are condemned properties
- ~26.9% of properties are dead-end properties

Figure V

- Some unique parcel ID's have more than 1 entry, showing duplicate parcel id issues Figure VI
 - ~27.3% of the properties (872 properties) have no valid addresses

Data Cleaning

 Created a helper function (score_adjust(x)) to adjust score ranges from 0-62 to 1-4

Score 1: 0-16
Score 2: 16-32
Score 3: 32-47
Score 4: 47-63

2. List of duplicate **parcel_id**'s was **371** entries before fixing the score range, then it was **9** entries.

- Created a helper function (remove_duplicates(data)) to remove the remaining 9 duplicate parcel ID entries:
 - Removed duplicate entries by keeping latest date, latest time, or highest score

score

2

3

4

3137

23

27

Name: count, dtype: int64

```
Parcel ID: 0022F00119000000
Parcel ID: 0022F00119000000
                                                    parcel_id    new_date    new_time
                                                                                      score
      parcel_id new_date new_time
                                     score
                                             0022F00119000000 2020-09-10 00:00:00
0022F00119000000 2020-09-10 00:00:00
                                         4
0022F00119000000 2020-09-10 00:00:00
                                         3
                                             Parcel ID: 0043S00018000000
Parcel ID: 0043S00018000000
                                                    parcel_id
                                                                new_date new_time
                                                                                      score
      parcel_id    new_date    new_time
                                     score
                                             0043500018000000 2022-07-27 00:00:00
0043500018000000 2020-08-11 00:00:00
                                         3
0043S00018000000 2022-07-27 00:00:00
```

Visualizations on ALL Properties

Figure I: Distribution new score range (1-4) of 2,802 properties

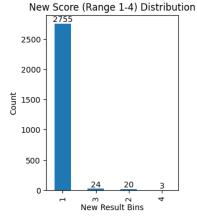


Figure II: Distribution of Parcel ID's

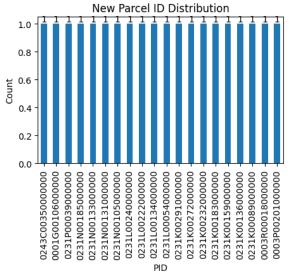
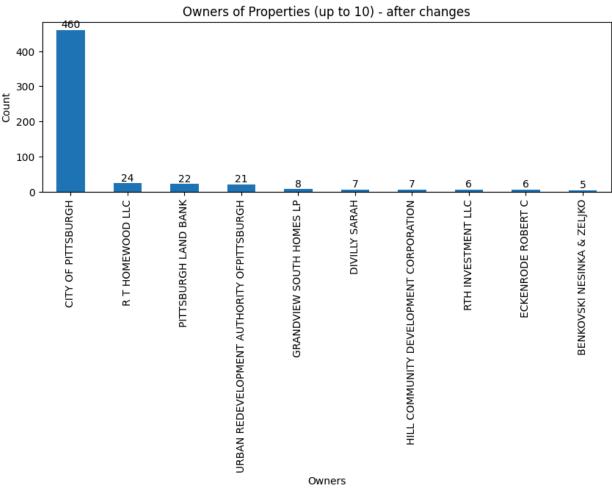


Figure III: Distribution of Owners of remaining properties



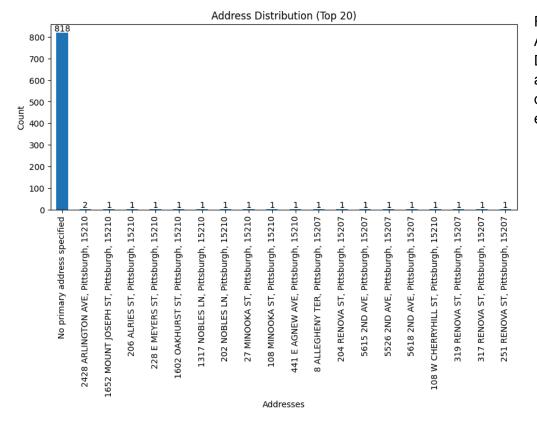


Figure IV:
Address
Distribution
after removing
duplicate
entries

Insights on ALL Properties

Figure I (new score range distribution)

- ~98.3% of the scores (2755 properties) are score 1, structurally intact with no immediate observable danger.
- ~0.7% of the scores (20 properties) are score 2, structurally compromised, unsafe, and possibly dangerous.
- ~0.9% of the scores (24 properties) are score 3, structurally compromised, unsafe, and dangerous.
- ~0.1% of the scores (3 properties) are score 4, imminently dangerous.

Figure II (new distribution of parcel ID's)

All Parcel IDs do not have any duplicates anymore

Figure III

- \bullet Now ~16.4% of properties are owned by the City of Pittsburgh, reduced from 503 to 460 Figure IV
 - ~29.2% of the properties have no valid addresses, reducing 872 to 818 properties

Condemned Properties

Visualizations on Condemned Properties (1,991 Properties)

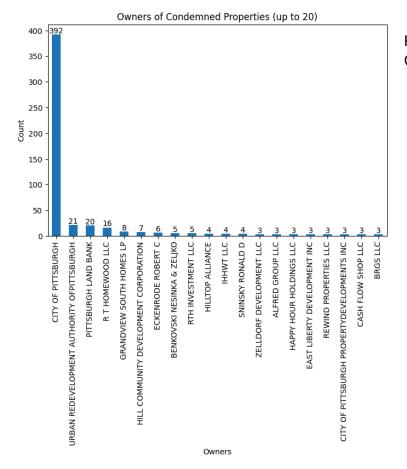


Figure I: Owners of the Condemned Properties

Figure II: Distribution of scores for Condemned Properties

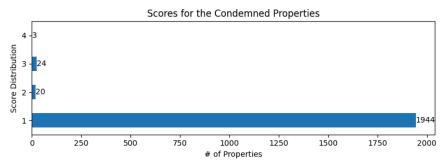


Figure III: Owners of the Score 4 Properties

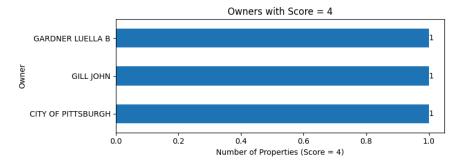


Figure IV: Pass or Fail results for Condemned Properties

Results of Condemned Properties (Pass/Fail)

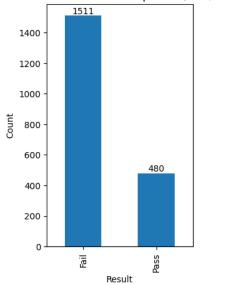
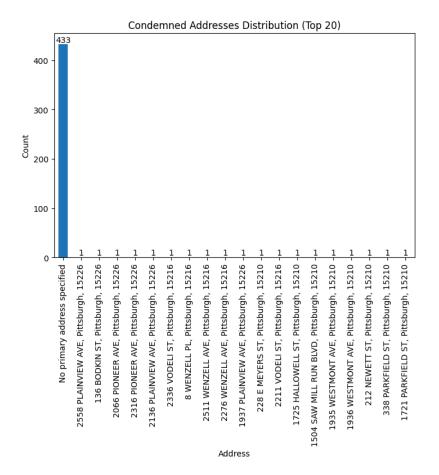


Figure V: Address distribution for Condemned Properties



Insights on Condemned Properties

Figure I

- ~97.6% of condemned properties have a score of 1.
- All 3 of the imminently dangerous properties (~0.2%) is condemned

Figure II

~19.7% (392) of the Condemned Properties are owned by City of Pittsburgh

Figure III

 The 3 owners of the dangerous properties are Gardner Luella B., Gill John, and the City of Pittsburgh

Figure IV

- ~75.9% of the Condemned Properties Fail the Inspection
- ~24.1% of the Condemned Properties Pass the Inspection

Figure V

• ~21.7% (433) of the Condemned Properties have no valid addresses.

Dead-End Properties

Visualizations on Dead-End Properties

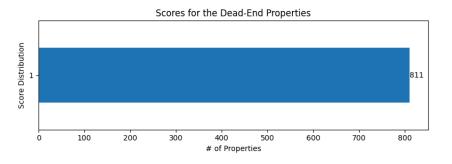
Figure I: Score Distribution for

Dead-End Properties

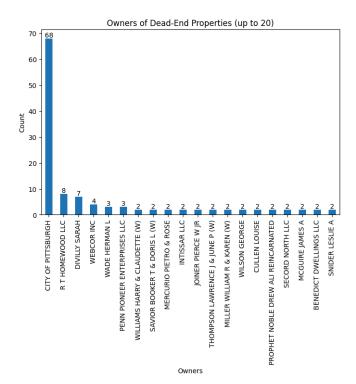
Figure II: Distribution of owners

of

Figure III: Pass or Fail for Dead-End Properties



Dead-End Properties



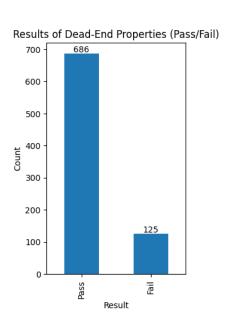
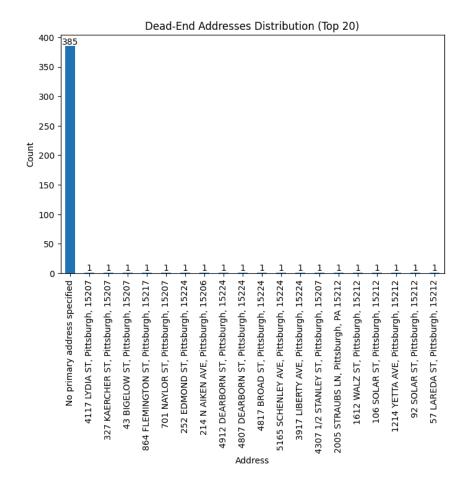


Figure IV: Dead-End Address Distribution



Insights on Dead-End Properties

Figure I

 100% of the Dead-End Properties are Score 1's, structurally intact with no immediate danger

Figure II

• ~8.4% of the Dead-End Properties are owned by City of Pittsburgh

Figure III

- ~84.6% of Dead-End Properties Fail the Inspections
- ~15.4% of Dead-End Properties Pass the Inspections

Figure IV

• ~47.5% of Dead-End Properties has no valid address

Overall Insights (from IPYNB)

Condemned or Dead-End

- Approximately 79.2% of Properties are Condemned
- · Approximately 20.8% of Properties are Dead-End

Fail Inspection

- Approximately 79.3% of Condemned Properties Fail Inspection.
- Approximately 10.6% of Dead-End Properties Fail Inspection.

Scores Evaluation

- All Dead-End Properties having a score of 1 show that they are structurally intact with no immediate observable danger. <u>Source</u>
- Condemned Properties scores range from 1-4, with 4 being imminently dangerous, these can be the properties/areas the city prioritizes.

Score 4 Properties

- The 3 owners of the 3 Score 4 Properties are: City of Pittsburgh, Gill John, and Gardner Luella B.
- By Identifying their Addresses with the cleaned dataset, we could prioritize these high-risk properties to obtain private demolition permit or a building permit to repair.

Addresses

- · Most addresses in the dataset is listed as "No primary address specified".
- This will have to be filled in to figure out which areas should be prioritized first.

After Merging with Allegheny Data

Process

- - Properties features will be combined to create a new Address feature using helper function build_address(row)
 - If address in Address feature is invalid, default to No primary address specified"
 - By matching the PARID with the parcel_id from original data frame using inner join, add Address column to the original data frame
 - Using helper function final_add(dataframe), address will be replaced with the valid address to new feature of final_address

Visualizations

Figure I: Address Distribution after adding Allegheny Data Addresses

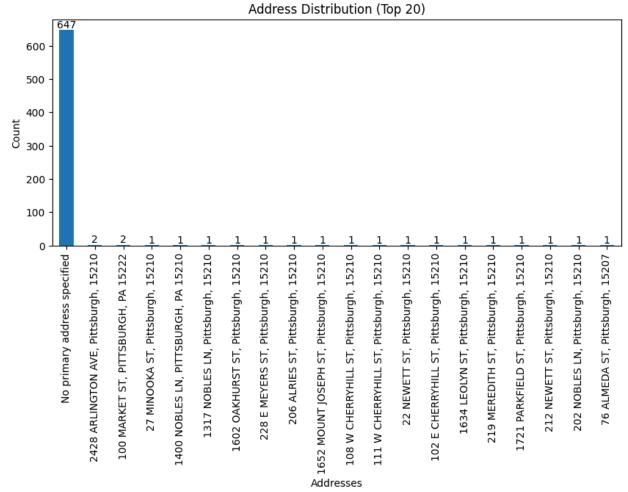
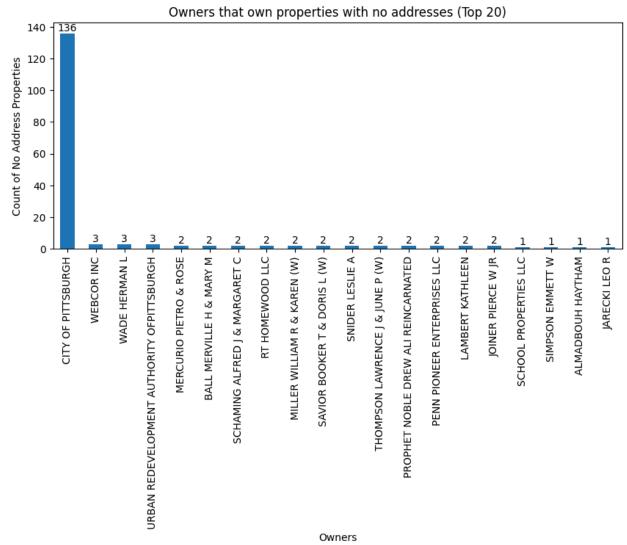


Figure II: Owners Distribution with no addresses



Insights

Figure I

Reduced No Address values from 813 to 647

Figure II

• City of Pittsburgh owns approximately 21% (136) of properties that do not have an address attached to it

After Merging with City Properties Data

Process

- 1. From the **City Properties Dataset** used features: ['pin', 'address', 'owner', 'acquisition date']
 - 'acquisition date' was converted into datetime data type to new feature date
 - 'owner' was renamed to city_owner for clarity
- 2. Inner join of the city data with matching parcel_id to pin
- 3. Created new feature **final_owner** by comparing **new_date** to **date** column
 - If new_date is latest, final_owner will take the owner value
 - If date is latest, final_owner will take the city_owner value
- 4. Created new feature **final_date** by comparing **new_date** to **date** column
 - If new_date is latest, final_date will take the new_date value
 - If date is latest, final_date will take the date value
- 5. Update **final_address** if final_address is not a valid address and new address from the city property dataset if valid
 - All 470 matching pid entries were updated with the owner, date, and address accordingly

Visualizations

Figure I:

parcel_id	address	owner	property_type	latest_inspection_result	inspection_status	new_date	new_time	score
0022F00119000000	No primary address specified	CITY OF PITTSBURGH	Condemned Property	Fail	Inactive	2020-09-10	00:00:00	4
0047E00052000000	No primary address specified	GILL JOHN	Condemned Property	Fail	Inactive	2020-09-04	00:00:00	4
0232A00021000000	No primary address specified	GARDNER LUELLA B	Condemned Property	Fail	Inactive	2020-07-10	00:00:00	4

Figure II:

parcel_id	owner	property_type	latest_inspection_result	inspection_status	new_date	new_time	score	final_address
0022F00119000000	CITY OF PITTSBURGH	Condemned Property	Fail	Inactive	2020-09-10	00:00:00		1406 WARNER ST, PITTSBURGH, PA 15233
0047E00052000000	GILL JOHN	Condemned Property	Fail	Inactive	2020-09-04 00:00:00	00:00:00		913 MINA ST, PITTSBURGH, PA 15212
0232A00021000000	GARDNER LUELLA B	Condemned Property	Fail	Inactive	2020-07-10 00:00:00	00:00:00	4	No primary address specified

Figure III:

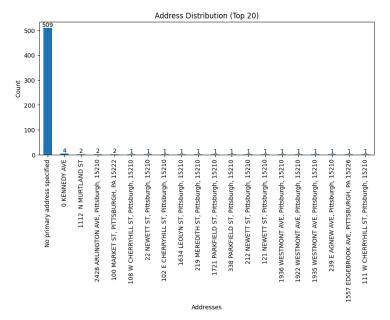


Figure IV:

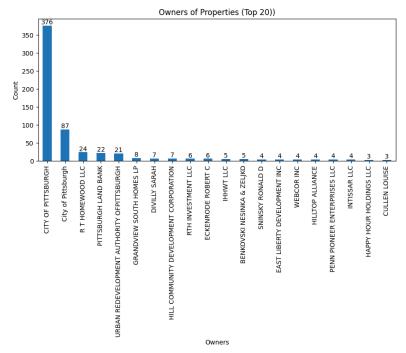
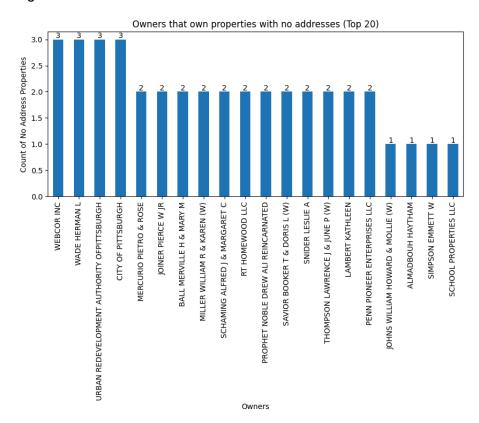


Figure V:



Insights

No Figure Dataset

- In our original cleaned data, City of Pittsburgh owned 460 properties
- The dataset claims that 470 of the properties are owned by the City of Pittsburgh
- 28 of the properties are claimed to be owned by the City of Pittsburgh when listed to be owned by another individual.

```
28 parcel id in city_df but not in city_props_df:
{'0045E00331000000', '0057G00191000000', '0014A00262000000', '0046N00197000000', '0026J00075000000'
```

- Filled in all 138 entries with matching ID's from City Property dataset that matched up with our parcel ID to update the 'No primary address specified' out of 470 entries.
- 87 entries' out of the 470 owners were updated based on the date of the entry.
- Lowered 647 missing address entries to 509. (saved in final_draft_df.csv) → this is difference of 647 and 138

Figures Insights

Figure I + II

- The 3 Score 4 Properties were all missing valid addresses
- After all addresses from Allegheny and City Properties were added, we are only missing one now
- Since the inspection status is inactive, we could push for obtaining private demolition permit or a building permit to repair

Figure III

- ~18.2% (509) of the address is still not valid, lowered from ~27.3% (872)
- 363 properties have been updated to a valid address, 138 properties updated from after Allegheny dataset

Figure IV

- ~13.4% (376) of the properties are owned by City of Pittsburgh, lowered from ~16.4% initial numbers (503)
- Means some of the owners were updated from the script replacing the owner with the latest entry

Figure VI

All properties owned by City of Pittsburgh was updated to valid address