Raw House Data Cleanse

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Dataset Description

5,000 observations regarding the qualities and cost of different houses

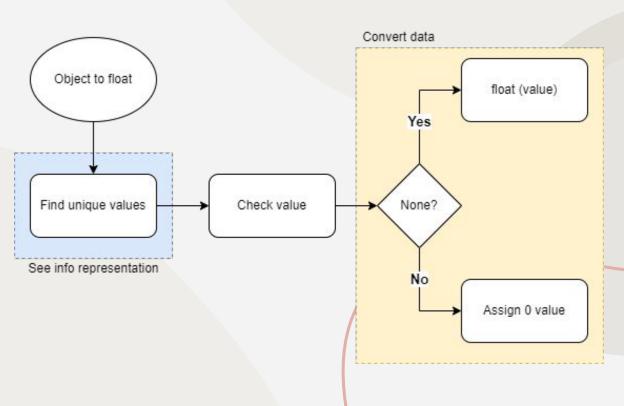
	Int	Float	Categorical
MLS	X		
sold_price		X	
zipcode	X		
longitude		X	
latitude		X	
lot_acres		X	
taxes		X	
year_built	X		

	Int	Float	Categorical
bedrooms	X		
bathrooms			X
sqrt_ft			X
garage			X
kitchen_features			X
fireplaces		X	
floor_covering			X
HOA			X

Data Type Conversion

Procedure for bathrooms, sqrt_ft, garage, and HOA

It was considered since there are other features with null values.



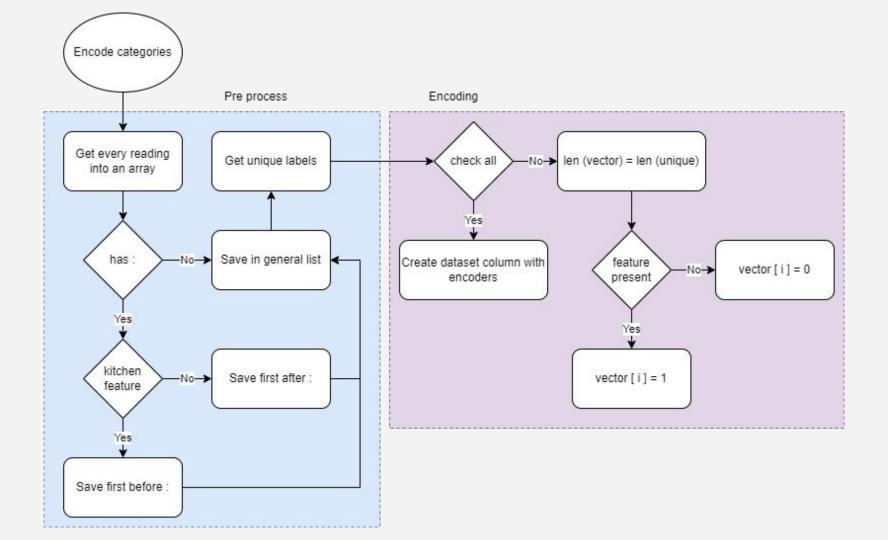
Encoding Categorical Features

Procedure made for kitchen_features and flor_covering.

Used multi-class encoding to get a single vector with every characteristic present for each house.

Created two new columns in the dataset:

- kitchen_vectors = 55-D vector
- floor_vectors = 87-D vector

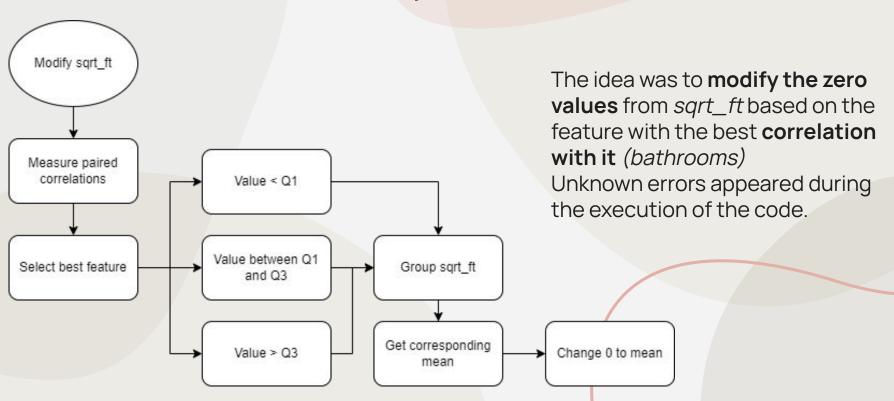


NaN and Outliers

Used EDA and boxplots to determine outliers

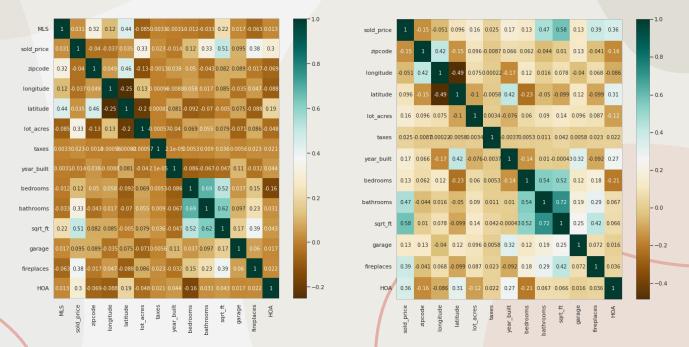
- Deleted null values from lot_acres and fireplaces
- Deleted column MLS (every value was unique)
- Used IQR method for deleting outliers in longitude, latitude and fireplaces
- Deleted taxes with 0 value and bigger than its Q3 score
- Deleted sqrt_ft with 0 value
- Deleted year_built before the 1800s
- Deleted sold_price bigger than 3M
- Kept zipcodes between 85400 and 86000
- Deleted lot_acres bigger than 500

Failed Implementations



Conclusions

- Deleted a total of 373 observations (7.46%) from the original dataset
- Included encoded representation of the two categorical variables
- A better correlation of features can be observed after the cleanse of data



Thank You!

Any Questions?