CSE 4510/5310 - Mgmt & Proc Big Data

Fall 2022

Hands-on Activity 1: Pandas - Exploratory Data Analysis (EDA)

Total Points: 40

Date Assigned: Friday, Sept 16, 2022

Due Date: Sunday, Sept 18, 2022

Submission Instructions: Please submit your work on Canvas as a Jupyter Notebook ipynb file named cse4510_yourname_activity1.ipynb

Key Big Data Processing Methods Demonstrated

- Preprocessing data
 - Loading CSV data using Pandas
 - Changing column names
 - Viewing snapshots of data
 - Converting data to different data types
 - Replacing strings in data and in field names
 - Plotting histograms from data
 - Plotting scatter plots to spot outliers in data
 - Removing outliers from data
 - Working with subsets of data
 - Plotting bar-charts from data

Download the file *rollingsales_brooklyn.csv* from Canvas. This dataset represents a selection of home sales made by RealDirect in Brooklyn, NY. RealDirect hires a team of licensed real-estate agents who work together and pool their knowledge to help people sell their homes. To accomplish this, it built an interface for sellers, giving them useful data-driven tips on how to sell their house. It also uses interaction data to give real-time recommendations on what to do next. RealDirect makes money by offering a subscription to sellers—about \$395 a month—to access the selling tools. Second, it allows sellers to use RealDirect's agents at a reduced commission, typically 2% of the sale instead of the usual 2.5% or 3%.

Imagine that you have been hired as chief data scientist (or analyst) at realdirect.com, and report directly to the CEO, Doug Perlson. The company (hypothetically) does not yet have its data plan in place.

Your task is to load in and clean up the rolling sales data. Next, conduct exploratory data analysis in order to find out where there are outliers or missing values, decide how you will treat them, make sure the dates are formatted correctly, make sure values you think are numerical are being treated as such, etc.

- 1. Reading the Data (4pts)
 - (a) Load the dataset using pandas
 - (b) Display the shape of the data to make sure it was read
 - (c) Display the first 2 rows
 - (d) Display row 14 to see the field names clearly



Figure 1: dtypes after completing 1d

- 2. Reformatting column labels (4pts)
 - (a) Convert the column names to lower case
 - (b) Replace the '\n' in the column names with space
 - (c) Replace the spaces in the column names with underscores
 - (d) Display row 14 again to see the change in the field names



Figure 2: row 14 after completing 2d

- 3. Convert the field "sale_price" to numeric (it is currently formatted as currency which cannot be used in calculations.) (Hints: There is a to_numeric function. You will first need to remove the commas and dollar signs. See slide 33 or 35 of the Pandas preprocessing tutorial.) (2pts)
- 4. Display a count of missing sale prices (those set to '0s') (1pt)

- 5. Convert "land_square_feet" to numeric (it's currently a formatted string which cannot be used in calculations) (1pt)
- 6. Convert "gross_square_feet" to numeric, "sale_date" to datetime, "year_built" to numeric (int32), and "zip_code" to string (4pts)

brooklyn_sales.dtypes	
borough	int64
neighborhood	object
building_class_category	object
tax_class_at_present	object
block	int64
lot	int64
ease-ment	object
building_class_at_present	object
address	object
apart_ment_number	object
zip_code	object
residential_units	int64
commercial_units	int64
total_units	int64
land_square_feet	int64
gross_square_feet	int64
year_built	int32
tax_class_at_time_of_sale	int64
<pre>building_class_at_time_of_sale</pre>	object
sale_price	int64
sale_date	datetime64[ns]
dtype: object	

Figure 3: dtypes after completing question 6

7. Change EAST/WEST to E/W for each address (2pts)

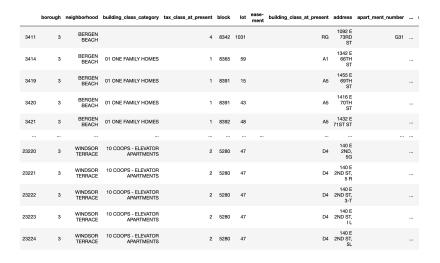


Figure 4: snapshot of the data after replacing EAST with E and WEST with W

8. Change STREET/AVENUE to ST/AVE for each address (2pts)

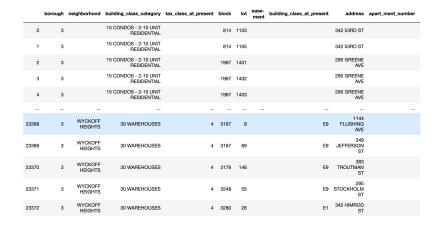


Figure 5: snapshot of the data after replacing STREET with ST and AVENUE with AVE

- 9. Plot the following histograms: (4pts)
 - sale price by counts
 - sale price>0
 - gross sqft where sale price == 0

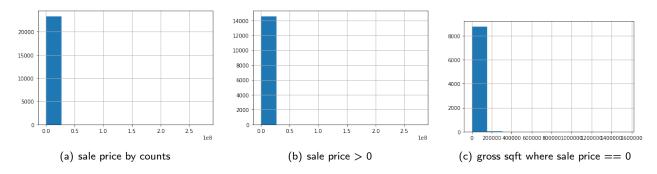


Figure 6: Question 9 charts

Working with a subset of the data

- 10. Filter the data by family homes (i.e., return a subset of rows in a temp variable where the "building_class_category" contains "FAMILY") (2pts)
- 11. Create a log scatter plot of "gross sqft" vs "sale price" for family homes (2pts)
- 12. Return a subset of family homes where sale price<100000. How many homes fall in this category? (2pts)
- 13. Return a count of the sale prices (Notice that some homes were sold for a \$1, etc. These seem to be outliers) (1pt)
- 14. Remove outliers of sales with sale price <= 5 (2pts)
- 15. With the outliers removed, create a plot of "gross sqft" vs "sale price" (2pts)

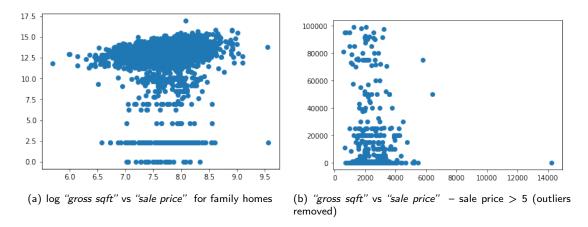


Figure 7: Charts for question 11 and 15

16. Create a plot of the top 10 home sales for family homes. (5pts)

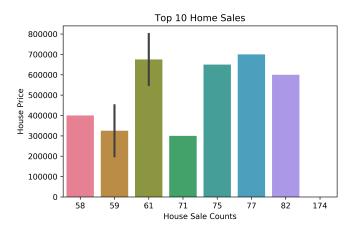


Figure 8: Chart for question 16

Bonus: Look Ma, I can make a meme with just code! (5pts)

Given the following 2 images, use only numpy and the PIL module to recreate the subsequent meme. Do not use loops. Each image in the frame is bordered with a red border using the code demonstrated in class. (Hint: Use the following resources to display text in PIL: https://pillow.readthedocs.io/en/stable/reference/ImageDraw.html, https://code-maven.com/create-images-with-python-pil-pillow)





Figure 9: Resulting Meme