NJRealtors_ML_Project

March 4, 2024

#Looking at the Big Picture ##### The first task is to use the New Jersey Realtor's state data to build a model that can predict median housing prices in each municipality using metrics sunch as New Listings, Closed Sales, Days on Market, Percent of Listing Price Received, Inventory of Homes, Months of Supply.

1 Framing the Problem

Objective: This model's output will be used to feed two other machine learning systems. One will use the predicted median sales price to determine the ARV of a renovated home. The other will determine if it's worth investing in the particular area/municipality

Current Solutions: The current solutions are to rely on industry experts (real estate agents and brokers) to manually create Comparable Market Analysis (CMA) which compares the target house to recently solds homes within a 0.5-1 mile radius which have similar property attributes. This method is highly unreliable because there currently isnt an industry standard as how to how to conduct a CMA. Agents are tasked to learn how to conduct this analysis either from other agents or learn by themselves; it is possible that no two agents will produce a similar CMA for the same property because their methodology will be different. In addition to the unreliability of the analysis, it is a very time consuming task and would benefit from being automated by a model.

Another solution available is the Zillow Zestimate price. Agents and or consumers will usually use the Zestimate as a trusted source in gauging the value of a target property. However, the model used to create the Zestimate is unknown and we arent certain on how that value is reached, although we are certain that machine learning/deep learning is involved.

2 Designing the System

My initial bias is to using a multiple regression model because it seems intuitive to use, however, I will approach this project with a true scientists mind. Coming in with zero assumptions and learning as I analyze different models, I will be exploring: #####1) If the data shows any significant correlation to warrant a machine learning project #####2) The use of various supervised and unsupervised models which tasks are used to solved reggression problems if there is significant correlation found.

#####This model will learn through batch learning

```
[62]: from sklearn.linear_model import LinearRegression
     from sklearn.preprocessing import StandardScaler
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import FunctionTransformer
     from sklearn.compose import TransformedTargetRegressor
     from sklearn.compose import ColumnTransformer
     from sklearn.pipeline import make pipeline
     from sklearn.metrics import mean_squared_error
     from sklearn.metrics import r2_score
     import math
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     from google.colab import drive
     import os
     import seaborn as sns
[63]: # /content is the default directory of Google Colab. Google Drive needs to be
      →mounted in order to access the My Drive directory
     current_wd = os.getcwd()
     drive.mount('/content/drive')
     Drive already mounted at /content/drive; to attempt to forcibly remount, call
     drive.mount("/content/drive", force_remount=True).
[64]: os.chdir('/content/drive/MyDrive/Colab Notebooks')
[65]: data = pd.read_excel('NJ 10k Real Estate Data 2024-01-25.xlsx', sheet_name='All_u
       data.head()
[65]:
                       Dates
                                       County Quarter
                                                           Month Year \
     City
     Aberdeen Twp 2019-12-01 Monmouth County
                                                   Q4
                                                        December
                                                                  2019
     Aberdeen Twp 2019-11-01 Monmouth County
                                                   Q4
                                                       November
                                                                  2019
     Aberdeen Twp 2019-10-01 Monmouth County
                                                   Q4
                                                         October
                                                                  2019
     Aberdeen Twp 2019-09-01 Monmouth County
                                                   Q3
                                                       September
                                                                  2019
     Absecon City 2019-12-01 Atlantic County
                                                        December
                                                                  2019
                                                   Q4
                   New Listings New Listing % Change (YoY) Closed Sales \
     City
     Aberdeen Twp
                                                                        9
                              6
                                                     -0.625
     Aberdeen Twp
                             12
                                                     -0.250
                                                                       21
     Aberdeen Twp
                             13
                                                     -0.235
                                                                       11
     Aberdeen Twp
                             19
                                                     -0.136
                                                                       15
```

Absecon City	11	0.100	18
City	Closed Sale % Change (YoY)	Days on Markets \	
Aberdeen Twp	-0.250	40	
Aberdeen Twp	0.313	48	
Aberdeen Twp	-0.450	84	
Aberdeen Twp	-0.318	56	
Absecon City	0.636	83	
City	Days on Market % Change (Yo	Y) Median Sales Prices	\
Aberdeen Twp	-0.1	342500	
Aberdeen Twp	-0.5		
Aberdeen Twp	0.5	380000	
Aberdeen Twp	0.3	376000	
Absecon City	0.2	297 190000	
Ci+v	Median Sales Price % Change	e (YoY) \	
City Aberdeen Twp		0.000	
Aberdeen Twp		0.043	
Aberdeen Twp		0.027	
Aberdeen Twp		0.153	
Absecon City		-0.050	
City	Percent of Listing Price Re	eceived \	
Aberdeen Twp		0.992	
Aberdeen Twp		0.973	
Aberdeen Twp		0.973	
Aberdeen Twp		0.974	
Absecon City		0.962	
City	Percent of Listing Price Re	eceive % Change (YoY) \	
Aberdeen Twp		-0.001	
Aberdeen Twp		0.003	
Aberdeen Twp		-0.028	
Aberdeen Twp		-0.035	
Absecon City		-0.003	
City	Inventory of Homes for Sale	es \	
Aberdeen Twp		31	
Aberdeen Twp		38	
Aberdeen Twp		39	

```
Aberdeen Twp
                                              50
                                              85
      Absecon City
                    Inventory of Homes for Sale % Change (YoY) Months of Supply \
      City
                                                        -0.295
      Aberdeen Twp
                                                                              2.1
      Aberdeen Twp
                                                        -0.240
                                                                              2.4
                                                                              2.5
      Aberdeen Twp
                                                        -0.188
      Aberdeen Twp
                                                                              3.1
                                                        -0.138
      Absecon City
                                                        -0.158
                                                                              6.2
                    Months of Supplies % Change (YoY)
      City
      Aberdeen Twp
                                               -0.276
                                               -0.314
      Aberdeen Twp
      Aberdeen Twp
                                               -0.242
      Aberdeen Twp
                                               -0.244
      Absecon City
                                               -0.151
[66]: # Will be using this dataset going forward to train all models
      X = data[['Median Sales Prices', 'New Listings', 'Closed Sales', 'Days on_
       GMarkets', 'Percent of Listing Price Received', 'Inventory of Homes for⊔
       ⇔Sales', 'Months of Supply']]
      X = X[data[['Median Sales Prices', 'New Listings', 'Closed Sales', 'Days on
       ⇔Markets', 'Percent of Listing Price Received', 'Inventory of Homes for⊔
      ⇔Sales', 'Months of Supply']] != 0]
      X = X.dropna()
      X.head()
[66]:
                    Median Sales Prices New Listings Closed Sales \
      City
      Aberdeen Twp
                               342500.0
                                                  6.0
                                                                9.0
      Aberdeen Twp
                               355000.0
                                                 12.0
                                                                21.0
                                                 13.0
      Aberdeen Twp
                               380000.0
                                                                11.0
      Aberdeen Twp
                               376000.0
                                                 19.0
                                                                15.0
      Absecon City
                               190000.0
                                                 11.0
                                                                18.0
                    Days on Markets Percent of Listing Price Received \
      City
                               40.0
                                                                  0.992
      Aberdeen Twp
      Aberdeen Twp
                               48.0
                                                                  0.973
      Aberdeen Twp
                               84.0
                                                                  0.973
      Aberdeen Twp
                               56.0
                                                                  0.974
      Absecon City
                               83.0
                                                                  0.962
                    Inventory of Homes for Sales Months of Supply
      City
```

```
      Aberdeen Twp
      31.0
      2.1

      Aberdeen Twp
      38.0
      2.4

      Aberdeen Twp
      39.0
      2.5

      Aberdeen Twp
      50.0
      3.1

      Absecon City
      85.0
      6.2
```

[67]: # These features are on completely different scales and will require.

standardization/normalization

X.describe().round(2)

[67]:		Median	Sales Prices	New	Listings	Closed S	ales	Days	on l	Markets	\
	count		24633.00		24633.00	2463	3.00		24	4633.00	
	mean		552197.76		15.50	1	2.68			44.84	
	std		448864.89		16.57	1	3.69			33.95	
	min		10000.00		1.00		1.00			1.00	
	25%		317500.00		5.00		4.00			23.00	
	50%		449500.00		10.00		8.00			36.00	
	75%		636250.00		20.00	1	6.00			56.00	
	max		12750000.00		205.00	20	9.00			585.00	
		Percent	of Listing	Price	Received	Inventor	y of	Homes	for	Sales	\
	count		· ·		24633.00				246	33.00	
	mean				1.01					34.84	
	std				0.05					45.04	
	min				0.57					1.00	
	25%				0.98					10.00	

1.00

1.03

1.81

20.00

40.00 737.00

	Months	of	Supply
count		24	4633.00
mean			2.82
std			2.24
min			0.10
25%			1.50
50%			2.20
75%			3.40
max			39.50

50%

75%

max

3 Correlation matrix for the target data set

3.0.1 Correlation: Unit of measurement of how two variables change with respect to one another

Correlations worth noting:

1. Median Sales Prices: Weak correlations with other features. 0.12 correlation with Months of

- Supply
- 2. New Listings: Very strong positive correlation (0.82) with Closed Sales and Inventory of Homes for Sales (0.67)
- 3. Closed Sales: Strong correlation with Inventory of Homes for Sale (0.63)
- 4. Days on Markets: Semi-strong negative correlation with Percent of Listong Prices Received (-0.425) and Months of Supply (0.27)
- 5. Percent of Listing Prices Received: Semi-strong negative correlation with Months of Supply (-0.296)

[68]: X.corr() [68]: Median Sales Prices New Listings \ Median Sales Prices 1.000000 -0.081203 New Listings -0.081203 1.000000 Closed Sales -0.089761 0.815420 Days on Markets 0.092728 -0.048853Percent of Listing Price Received -0.081821 0.089906 Inventory of Homes for Sales -0.067601 0.659787 Months of Supply 0.122587 0.023898 Closed Sales Days on Markets Median Sales Prices -0.089761 0.092728 New Listings 0.815420 -0.048853 Closed Sales 1.000000 -0.082559 Days on Markets -0.082559 1.000000 Percent of Listing Price Received 0.106724 -0.441308 Inventory of Homes for Sales 0.623398 0.092321 Months of Supply -0.085192 0.291219 Percent of Listing Price Received Median Sales Prices -0.081821 New Listings 0.089906 Closed Sales 0.106724 Days on Markets -0.441308 Percent of Listing Price Received 1.000000 Inventory of Homes for Sales -0.099902 Months of Supply -0.309631 Inventory of Homes for Sales \ Median Sales Prices -0.067601 New Listings 0.659787 Closed Sales 0.623398 Days on Markets 0.092321 Percent of Listing Price Received -0.099902 Inventory of Homes for Sales 1.000000 Months of Supply 0.465619

	Months of Supply
Median Sales Prices	0.122587
New Listings	0.023898
Closed Sales	-0.085192
Days on Markets	0.291219
Percent of Listing Price Received	-0.309631
Inventory of Homes for Sales	0.465619
Months of Supply	1.000000

4 Covariance

4.0.1 Covariance: A measure of how two variables move together with respect to each other

Additional Info:

- 1. Helpful to determine the type of relationship between variables
- 2. Magnitude not easily interpretable
- 3. Higher magnitude means more dependent relationship. An increase in one variable would lead to an increase in another variable
- 4. A negative covaraince denotes an inverse relationship between the two varaibles

[71]: X.cov()

	Median Sales P	rices N	New Listin	gs
Median Sales Prices	2.01479	7e+11 -60	3785.4302	47
New Listings	-6.03785	4e+05	274.4026	56
Closed Sales	-5.51693	2e+05	184.9568	65
Days on Markets	1.41294	1e+06	-27.4713	27
Percent of Listing Price Received	-1.72714	5e+03	0.0700	37
Inventory of Homes for Sales	-1.36671	0e+06	492.2713	04
Months of Supply	1.23380	1e+05	0.8876	64
	Closed Sales	Days on	Markets	\
Median Sales Prices	-551693.162358	1.412	2941e+06	
New Listings	184.956865	-2.747	7133e+01	
Closed Sales	187.495261	-3.837	7599e+01	
Days on Markets	-38.375994	1.152	2382e+03	
Percent of Listing Price Received	0.068723	-7.045	5098e-01	
Inventory of Homes for Sales	384.474432	1.411	l587e+02	
Months of Supply	-2.615656	2.216	6676e+01	
	Percent of Lis	ting Prid	ce Receive	d
Median Sales Prices		-1	1727.14481	6
New Listings			0.07003	7
Closed Sales			0.06872	3
Days on Markets			-0.70451	0

Percent of Listing Price Received	0.002212
Inventory of Homes for Sales	-0.211607
Months of Supply	-0.032649

	Inventory of	Homes for Sales	\
Median Sales Prices		-1.366710e+06	
New Listings		4.922713e+02	
Closed Sales		3.844744e+02	
Days on Markets		1.411587e+02	
Percent of Listing Price Received		-2.116067e-01	
Inventory of Homes for Sales		2.028680e+03	
Months of Supply		4.702427e+01	

	Months of Supply
Median Sales Prices	123380.139880
New Listings	0.887664
Closed Sales	-2.615656
Days on Markets	22.166764
Percent of Listing Price Received	-0.032649
Inventory of Homes for Sales	47.024273
Months of Supply	5.027694

5 Seaborn Paiplot Findings

There's a possible linear relationship between New Listings and Closed Sales but outside of that, there seems to be no linear relationships between the data. This may possibly mean that a multiple regression model may not produce accurate results.

```
[72]: sns.pairplot(X, y_vars='Median Sales Prices', hue='Median Sales Prices') plt.show()
```

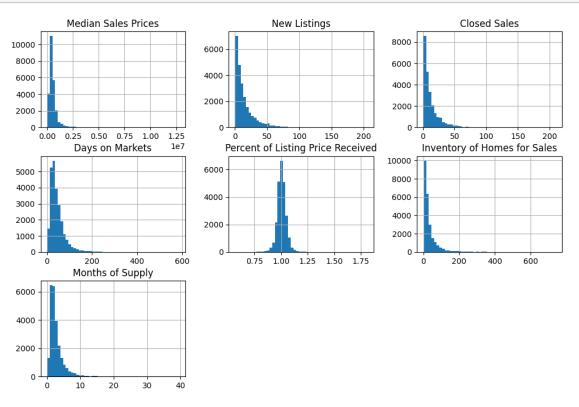


6 Histogram

The scale for the Median Sales Prices in the Seaborn and Matplotlib plot are heavily misaligned. Its on a 1e7 scale and ALL of the data is skewed right between 0 and 1.275e7 which is 12,750,000 which doesnt show an ideal visualization of the data.

The scale of all the data are not aligned are good candidates to be transformed and standardized

```
[73]: X.hist(bins=50, figsize=(12,8))
plt.show()
```



#Transformations

####After visualizing the relationships between the dependent and independent variables, as well as the distribution of the independent variables, it is clear that we need to implement some transformations to create a more normal distribution amongst the data and standardize the scales.

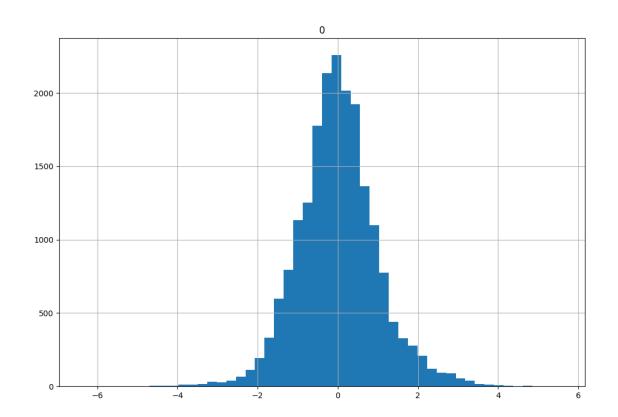
#####There needs to be a global transformation as well as a local transformation used on the data. The global transformation used will be Standardization. Currently, the scales of the data ranges from 0 - 200 for New Listings and Closed Sales, 0 - 600 for Days on the Market and Inventory of Homes for Sale, 0 - 40 for Months of Supply, 0 - 1.75 for Percent of Listing Price Received (PoLPR) and 0 - 1,250,000 for Median Sales Prices. The Standardization scaler will squash these values down to a smaller range but wont change the shape or the distribution of the data.

####However, before using the Standardization transformation, the data needs to be evenly distributed. Based on the pairplot and historgram plots, the only data that seems to be evenly distributed in the PoLPR. All other feature columns to have heavy tails (right or positively skewed). When dealing with right skewed data, it is best to use log-transforms to help make that distribution more even.

```
[74]: # Create a train, test, split on the data before initiating the transforms
y = X['Median Sales Prices']
x = X.drop(columns=['Median Sales Prices'])
```

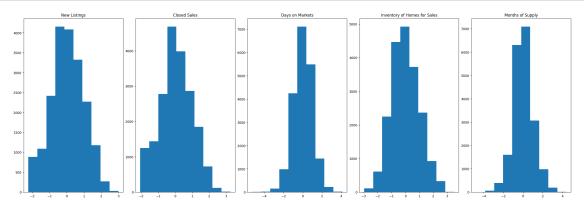
```
# It is industry standard to use an 80/20 split for the test and training data
      x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2,__
       →random_state=3)
      orginal_y_train = y_train.copy()
[75]: # Create a sklearn pipeline to apply the log-transform on the columns which are
      ⇔right-skewed then StandardScalar()
      # Use the FunctionTransformer method to apply the np.log transform
      # Use the ColumnTransformer() to properly apply the transforms to the target,
       ⇔columns based on what it needs
      standarizer = StandardScaler()
      log_transform = FunctionTransformer(np.log, inverse_func=np.exp)
      log_pipeline = make_pipeline(log_transform, standarizer)
      default_pipeline = make_pipeline(standarizer)
      log_target_features = ['New Listings', 'Closed Sales', 'Days on Markets', |
       ⇔'Inventory of Homes for Sales', 'Months of Supply']
      default_target_features = ['Percent of Listing Price Received']
      pre_processing = ColumnTransformer([('log', log_pipeline, log_target_features),
                                          ('default', default_pipeline, __

default_target_features)])
[76]: def target_label_transformer(labels, inverse=False, standard_func=None):
        standarizer= StandardScaler()
        if inverse is False:
          trans_label = log_transform.fit_transform(labels)
          trans_label = standarizer.fit_transform(trans_label.to_frame())
          return trans_label, standarizer
        elif inverse is True:
          trans_label = standard_func.inverse_transform(labels)
          trans_label = np.exp(trans_label)
          return trans_label
     ###Standardized 'Median Sales Prices'
[77]: new_labels, standarizer_func = target_label_transformer(y_train, inverse=False)
      y_train = pd.DataFrame(new_labels)
      y_train.hist(bins=50, figsize=(12,8))
[77]: array([[<Axes: title={'center': '0'}>]], dtype=object)
```



6.0.1 Standardized Features

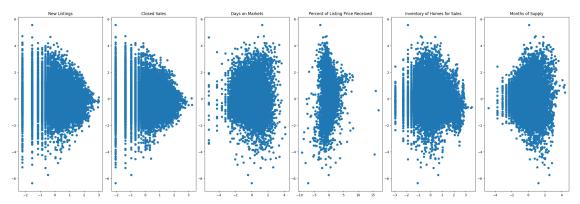
```
[78]: fig, axs = plt.subplots(ncols=5, figsize=(24,8), layout='constrained')
for idx, col in enumerate(log_target_features):
    #td means transformed data
    td, standarizer_func = target_label_transformer(x_train[col])
    axs[idx].hist(td)
    axs[idx].set_title(col)
```



###Scatterplots of 'Median Sales Prices' vs. other feature elements to check for correlation

```
[79]: target, standarizer_func = target_label_transformer(orginal_y_train)
    fig, axs = plt.subplots(ncols=6, figsize=(24,8), layout='constrained')
    for idx, col in enumerate(x_train.columns):
        #td means transformed data
        if col == 'Percent of Listing Price Received':
            td = StandardScaler().fit_transform(x_train[col].to_frame())
        else:
            td, standarizer_func = target_label_transformer(x_train[col])

        axs[idx].scatter(td, target)
        axs[idx].set_title(col)
```



```
[80]: # Create another pipeline which links the pre-processing var and the
       LinearRegression() method then train the model on the training data and
       ⇔score it
      x_prepared = pre_processing.fit_transform(x_train)
      # The target values need to be standardized as well
      training y, standarizer func1 = target_label_transformer(orginal_y_train,_
       →inverse=False)
      # model = TransformedTargetRegressor(LinearRegression(),__
      ⇔transformer=StandardScaler())
      model = LinearRegression()
      model.fit(x_prepared, training_y)
      y_hat = model.predict(pre_processing.fit_transform(x_test))
      rmse = math.sqrt(mean_squared_error(y_test, target_label_transformer(y_hat,_
       →inverse=True, standard_func=standarizer_func1)))
      print("Mean squared error: %.2f" % rmse)
      print("Coefficient of determination: %.2f" % r2_score(y_test,_
       →target_label_transformer(y_hat, inverse=True, __
       standard_func=standarizer_func1)))
```

```
# plt.scatter(x_test, pd.DataFrame(target_label_transformer(y_hat, u → inverse=True, standard_func=standarizer_func1)))
```

Mean squared error: 492639.38

Coefficient of determination: -0.05

###Results

####Multiple Linear Regression doesn't seem to be the best generalizing function to find the effects of different independent variables on the Median Sales Price. Even looking at each individual independent feature vs Median Sale Prices shows no indication of linear correlation between the variables. I believe this checks out well with the correlation matrix generated in the beginning of the project. I will need to choose a stronger model to determine any type of insights or trends

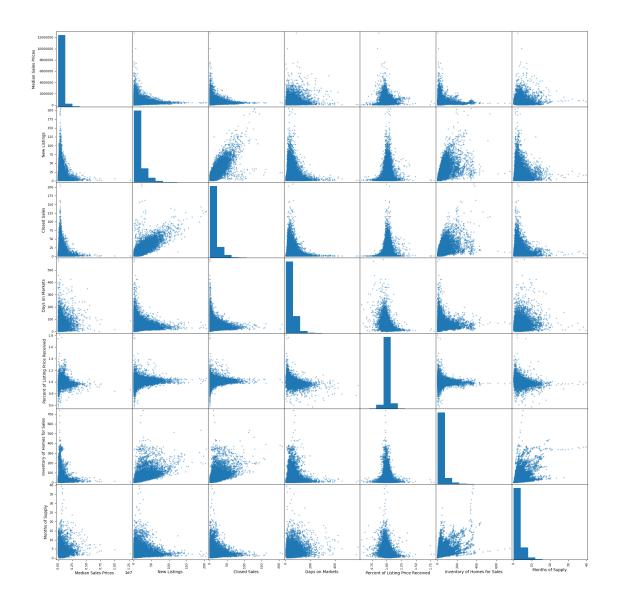
###Questions

#####1) Can I preview the correlation matrix to determine the relationship between the dependent and independent variables to determine if this is a good candidate for multiple linear regression? #####2) Can data have very little correlation but very high covariance? #####3) Would adding the month and year variables to the model make a difference? #####4) Are there any previous steps that were done incorrectly to lead me to these conclusions? #####5) Would using a regualrized(Ridge or Lasso) regression model make a difference? #####6) Is this bad data? Is there any feature engineering I can do to improve the data?

[81]: X.corr()

81]:		Median Sales	Prices	New Listings	\
	Median Sales Prices	1.	.000000	-0.081203	
	New Listings	-0.	.081203	1.000000	
	Closed Sales	-0.	.089761	0.815420	
	Days on Markets	0.	.092728	-0.048853	
	Percent of Listing Price Received	-0.	.081821	0.089906	
	Inventory of Homes for Sales	-0.	.067601	0.659787	
	Months of Supply	0.	.122587	0.023898	
		Closed Sales	•		
	Median Sales Prices	-0.089761		0.092728	
	New Listings	0.815420		-0.048853	
	Closed Sales	1.000000		-0.082559	
	Days on Markets	-0.082559		1.000000	
	Percent of Listing Price Received	0.106724		-0.441308	
	Inventory of Homes for Sales	0.623398		0.092321	
	Months of Supply	-0.085192		0.291219	
		Percent of Li	isting I	Price Received	\
	Median Sales Prices			-0.081821	
	New Listings			0.089906	
	Closed Sales			0.106724	

```
Days on Markets
                                                                 -0.441308
     Percent of Listing Price Received
                                                                  1.000000
      Inventory of Homes for Sales
                                                                 -0.099902
      Months of Supply
                                                                  -0.309631
                                         Inventory of Homes for Sales \
     Median Sales Prices
                                                            -0.067601
     New Listings
                                                             0.659787
     Closed Sales
                                                             0.623398
     Days on Markets
                                                             0.092321
     Percent of Listing Price Received
                                                            -0.099902
      Inventory of Homes for Sales
                                                             1.000000
     Months of Supply
                                                             0.465619
                                         Months of Supply
     Median Sales Prices
                                                 0.122587
     New Listings
                                                 0.023898
     Closed Sales
                                                -0.085192
     Days on Markets
                                                 0.291219
      Percent of Listing Price Received
                                                -0.309631
      Inventory of Homes for Sales
                                                 0.465619
     Months of Supply
                                                 1.000000
[82]: from pandas.plotting import scatter_matrix
      scatter_matrix(X, figsize=(24,24))
      plt.show()
```



6.0.2 Multiple Linear Regression: Feature Engineering

As a real estate agent, I know there are many factors which effect home prices; and one of those major factors are mortgage rates. I'm interested to see if adding 'Mortgage Rates' as a feature of the data will convey any correlation to the 'Median Home Prices'

```
new_data = new_data.dropna()
      modify dates = new data['Dates'].to list()
      modify_dates = [str(i)[:7] for i in modify_dates]
      new_data.drop(columns=['Dates'], inplace=True)
      new_data.insert(0, 'Dates', pd.to_datetime(modify_dates, format='%Y-%m'))
      new_data.set_index('Dates', inplace=True, drop=False)
      new_data['Mortgage Rates'] = 0
[84]: mortgage_data = pd.read_excel('Mortgage_Rates_1971 - 2024.xlsx')
      observation_dates = mortgage_data['observation_date'].to_list()
      observation_dates = [str(i)[:7] for i in observation_dates]
      mortgage_data.drop(columns=['observation_date'], inplace=True)
      mortgage_data.insert(0, 'Dates', pd.to_datetime(observation_dates,__

¬format="%Y-%m"))
[85]: def get_mortgage_rate(mortgage_df, main_df):
          mortgage df = mortgage df.loc[2515:]
          for idx, date in enumerate(set(mortgage_df['Dates'].to_list())):
              median_mortgage_rate = mortgage_df[mortgage_df['Dates'] == date]
              main_df.loc[date, 'Mortgage Rates'] =__
       →median_mortgage_rate['MORTGAGE30US_20240229'].mean()
          return main df
[86]: new_df = get_mortgage_rate(mortgage_data, new_data)
      new_df.set_index('Dates', inplace=True, drop=False)
      new_df.describe()
[86]:
             Median Sales Prices
                                  New Listings
                                                 Closed Sales
                                                               Days on Markets
      count
                    2.463300e+04
                                   24633.000000
                                                 24633.000000
                                                                   24633.000000
     mean
                    5.521978e+05
                                                                      44.838956
                                      15.500954
                                                    12.680997
      std
                    4.488649e+05
                                      16.565104
                                                    13.692891
                                                                      33.946754
     min
                    1.000000e+04
                                       1.000000
                                                     1.000000
                                                                       1.000000
      25%
                    3.175000e+05
                                       5.000000
                                                     4.000000
                                                                      23.000000
      50%
                    4.495000e+05
                                      10.000000
                                                     8.000000
                                                                      36.000000
      75%
                    6.362500e+05
                                      20.000000
                                                    16.000000
                                                                      56.000000
      max
                    1.275000e+07
                                     205.000000
                                                   209.000000
                                                                     585.000000
                                                 Inventory of Homes for Sales
             Percent of Listing Price Received
                                   24633.000000
      count
                                                                  24633.000000
                                       1.006126
                                                                     34.839808
     mean
      std
                                       0.047027
                                                                     45.040873
     min
                                       0.571000
                                                                      1.000000
      25%
                                       0.982000
                                                                     10.000000
      50%
                                       1.005000
                                                                     20,000000
      75%
                                       1.030000
                                                                     40.000000
                                       1.807000
                                                                    737.000000
     max
```

	Months of Supply	Mortgage Rates
count	24633.000000	24643.000000
mean	2.815508	4.508294
std	2.242252	1.684395
min	0.100000	2.684000
25%	1.500000	3.016000
50%	2.200000	3.688000
75%	3.400000	6.342500
max	39.500000	7.620000

[87]: new_df.corr()

<ipython-input-87-326e7bbec5b0>:1: FutureWarning: The default value of
numeric_only in DataFrame.corr is deprecated. In a future version, it will
default to False. Select only valid columns or specify the value of numeric_only
to silence this warning.

new_df.corr()

[87]:		Median Sales	Prices	New Listi	ngs	\
	Median Sales Prices	1.	000000	-0.081	203	
	New Listings	-0.	1.000000			
	Closed Sales	-0.	0.815	420		
	Days on Markets	0.	092728	-0.048	853	
	Percent of Listing Price Received	-0.	081821	0.089	906	
	Inventory of Homes for Sales	-0.	067601	0.659	787	
	Months of Supply	0.	122587	0.023	898	
	Mortgage Rates	0.	127868	-0.147	793	
			_			
		Closed Sales	•		\	
	Median Sales Prices	-0.089761		0.092728		
	New Listings	0.815420		-0.048853		
	Closed Sales	1.000000		-0.082559		
	Days on Markets	-0.082559		1.000000		
	Percent of Listing Price Received	0.106724		-0.441308		
	Inventory of Homes for Sales	0.623398		0.092321		
	Months of Supply	-0.085192		0.291219		
	Mortgage Rates	-0.153428		-0.164820		
		Percent of Li	sting D	Price Pecci	borr	\
	Median Sales Prices	reicent of Li	sting r	-0.081		`
	New Listings			0.089		
	Closed Sales			0.106		
	Days on Markets			-0.441		
	Percent of Listing Price Received			1.000		
	Inventory of Homes for Sales			-0.099		
	·			-0.099		
	Months of Supply			-0.309	031	

Mortgage Rates 0.178073

	Inventory	of	Homes	for	Sales	\
Median Sales Prices				-0.	067601	
New Listings				0.0	659787	
Closed Sales				0.0	623398	
Days on Markets				0.0	092321	
Percent of Listing Price Received				-0.	099902	
Inventory of Homes for Sales				1.0	000000	
Months of Supply				0.4	465619	
Mortgage Rates				-0.	103271	

	Months of Supply	y Mortgage Rates
Median Sales Prices	0.12258	0.127868
New Listings	0.023898	-0.147793
Closed Sales	-0.08519	2 -0.153428
Days on Markets	0.291219	-0.164820
Percent of Listing Price Received	-0.30963	0.178073
Inventory of Homes for Sales	0.465619	9 -0.103271
Months of Supply	1.00000	-0.014656
Mortgage Rates	-0.01465	1.000000

6.0.3 Conclusion

The additional feature of 'Mortgage Rates' didnt seem to have any material effect on the 'Median Sales Price' or any of the other features for that matter. While knowing that mortgage rates have an effect on the housing market in real life, it may be difficult to show it's relationship here. Additionally, this study leads me to believe that home prices are effected more by the intrinsic values of the home and other factors not captured in this data set. I've determined that multiple linear regression

Reading package lists... Done Building dependency tree... Done Reading state information... Done

The following additional packages will be installed:

dvisvgm fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre fonts-urw-base35 libapache-pom-java libcommons-logging-java libcommons-parent-java libfontbox-java libfontenc1 libgs9 libgs9-common libidn12 libijs-0.35 libjbig2dec0 libkpathsea6 libpdfbox-java libptexenc1 libruby3.0 libsynctex2 libteckit0 libtexlua53 libtexluajit2 libwoff1 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-net-telnet ruby-rubygems ruby-webrick ruby-xmlrpc ruby3.0 rubygems-integration tlutils teckit tex-common tex-gyre texlive-base

texlive-binaries texlive-latex-base texlive-latex-extra texlive-latex-recommended texlive-pictures tipa xfonts-encodings xfonts-utils

Suggested packages:

fonts-noto fonts-freefont-otf | fonts-freefont-ttf libavalon-framework-java libcommons-logging-java-doc libexcalibur-logkit-java liblog4j1.2-java poppler-utils ghostscript fonts-japanese-mincho | fonts-ipafont-mincho fonts-japanese-gothic | fonts-ipafont-gothic fonts-arphic-ukai fonts-arphic-uming fonts-nanum ri ruby-dev bundler debhelper gv | postscript-viewer perl-tk xpdf | pdf-viewer xzdec texlive-fonts-recommended-doc texlive-latex-base-doc python3-pygments icc-profiles libfile-which-perl libspreadsheet-parseexcel-perl texlive-latex-extra-doc texlive-latex-recommended-doc texlive-luatex texlive-pstricks dot2tex prerex texlive-pictures-doc vprerex default-jre-headless tipa-doc

The following NEW packages will be installed:

dvisvgm fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre fonts-urw-base35 libapache-pom-java libcommons-logging-java libcommons-parent-java libfontbox-java libfontenc1 libgs9 libgs9-common libidn12 libijs-0.35 libjbig2dec0 libkpathsea6 libpdfbox-java libptexenc1 libruby3.0 libsynctex2 libteckit0 libtexlua53 libtexluajit2 libwoff1 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-net-telnet ruby-rubygems ruby-webrick ruby-xmlrpc ruby3.0 rubygems-integration t1utils teckit tex-common tex-gyre texlive-base texlive-binaries texlive-fonts-recommended texlive-latex-base texlive-latex-extra texlive-latex-recommended texlive-pictures texlive-plain-generic texlive-xetex tipa xfonts-encodings xfonts-utils O upgraded, 54 newly installed, O to remove and 35 not upgraded.

Need to get 182 MB of archives.

After this operation, 571 MB of additional disk space will be used.

Get:1 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-droid-fallback all 1:6.0.1r16-1.1build1 [1,805 kB]

Get:2 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-lato all 2.0-2.1 [2,696 kB]

Get:3 http://archive.ubuntu.com/ubuntu jammy/main amd64 poppler-data all 0.4.11-1 [2,171 kB]

Get:4 http://archive.ubuntu.com/ubuntu jammy/universe amd64 tex-common all 6.17 [33.7 kB]

Get:5 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-urw-base35 all 20200910-1 [6,367 kB]

Get:6 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libgs9-common all 9.55.0~dfsg1-Oubuntu5.6 [751 kB]

Get:7 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libidn12 amd64 1.38-4ubuntu1 [60.0 kB]

Get:8 http://archive.ubuntu.com/ubuntu jammy/main amd64 libijs-0.35 amd64 0.35-15build2 [16.5 kB]

Get:9 http://archive.ubuntu.com/ubuntu jammy/main amd64 libjbig2dec0 amd64 0.19-3build2 [64.7 kB]

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Get:10 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libgs9 amd64 9.55.0~dfsg1-Oubuntu5.6 [5,031 kB]
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Get:11 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libkpathsea6 amd64 2021.20210626.59705-1ubuntu0.1 [60.3 kB]

Get:12 http://archive.ubuntu.com/ubuntu jammy/main amd64 libwoff1 amd64
1.0.2-1build4 [45.2 kB]

Get:13 http://archive.ubuntu.com/ubuntu jammy/universe amd64 dvisvgm amd64 2.13.1-1 [1,221 kB]

Get:14 http://archive.ubuntu.com/ubuntu jammy/universe amd64 fonts-lmodern all 2.004.5-6.1 [4,532 kB]

Get:15 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-noto-mono all 20201225-1build1 [397 kB]

Get:16 http://archive.ubuntu.com/ubuntu jammy/universe amd64 fonts-texgyre all 20180621-3.1 [10.2 MB]

Get:17 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libapache-pom-java all 18-1 [4,720 B]

Get:18 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libcommons-parent-java all 43-1 [10.8 kB]

Get:19 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libcommons-logging-java all 1.2-2 [60.3 kB]

Get:20 http://archive.ubuntu.com/ubuntu jammy/main amd64 libfontenc1 amd64 1:1.1.4-1build3 [14.7 kB]

Get:21 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libptexenc1 amd64 2021.20210626.59705-1ubuntu0.1 [39.1 kB]

Get:22 http://archive.ubuntu.com/ubuntu jammy/main amd64 rubygems-integration all 1.18 [5,336 B]

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3.0.2-7ubuntu2.4 [50.1 kB]

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3.3.5-2 [228 kB]

Get:25 http://archive.ubuntu.com/ubuntu jammy/main amd64 ruby amd64 1:3.0~exp1
[5,100 B]

Get:26 http://archive.ubuntu.com/ubuntu jammy/main amd64 rake all 13.0.6-2 [61.7 kB]

Get:27 http://archive.ubuntu.com/ubuntu jammy/main amd64 ruby-net-telnet all
0.1.1-2 [12.6 kB]

Get:28 http://archive.ubuntu.com/ubuntu jammy/universe amd64 ruby-webrick all
1.7.0-3 [51.8 kB]

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2.5.11+ds1-1 [421 kB]

Get:33 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libtexlua53 amd64 2021.20210626.59705-1ubuntu0.1 [120 kB]

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0.13.72+dfsg.1-1.1 [27.0 kB]

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Get:38 http://archive.ubuntu.com/ubuntu jammy/universe amd64 lmodern all 2.004.5-6.1 [9,471 kB]

Get:39 http://archive.ubuntu.com/ubuntu jammy/universe amd64 preview-latex-style all 12.2-1ubuntu1 [185 kB]

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1.41-4build2 [61.3 kB]

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20180621-3.1 [6,209 kB]

Get:43 http://archive.ubuntu.com/ubuntu jammy-updates/universe amd64 texlive-binaries amd64 2021.20210626.59705-1ubuntu0.1 [9,848 kB]

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1:1.8.16-2 [207 kB]

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1:1.8.16-2 [5,199 kB]

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Get:54 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-xetex all 2021.20220204-1 [12.4 MB]

Fetched 182 MB in 3s (72.0 MB/s)

debconf: unable to initialize frontend: Dialog

debconf: (No usable dialog-like program is installed, so the dialog based

frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line 78, <> line 54.)

debconf: falling back to frontend: Readline

```
debconf: unable to initialize frontend: Readline
debconf: (This frontend requires a controlling tty.)
debconf: falling back to frontend: Teletype
dpkg-preconfigure: unable to re-open stdin:
Selecting previously unselected package fonts-droid-fallback.
(Reading database ... 121749 files and directories currently installed.)
Preparing to unpack .../00-fonts-droid-fallback 1%3a6.0.1r16-1.1build1 all.deb
Unpacking fonts-droid-fallback (1:6.0.1r16-1.1build1) ...
Selecting previously unselected package fonts-lato.
Preparing to unpack .../01-fonts-lato_2.0-2.1_all.deb ...
Unpacking fonts-lato (2.0-2.1) ...
Selecting previously unselected package poppler-data.
Preparing to unpack .../02-poppler-data_0.4.11-1_all.deb ...
Unpacking poppler-data (0.4.11-1) ...
Selecting previously unselected package tex-common.
Preparing to unpack .../03-tex-common_6.17_all.deb ...
Unpacking tex-common (6.17) ...
Selecting previously unselected package fonts-urw-base35.
Preparing to unpack .../04-fonts-urw-base35 20200910-1 all.deb ...
Unpacking fonts-urw-base35 (20200910-1) ...
Selecting previously unselected package libgs9-common.
Preparing to unpack .../05-libgs9-common_9.55.0~dfsg1-0ubuntu5.6_all.deb ...
Unpacking libgs9-common (9.55.0~dfsg1-Oubuntu5.6) ...
Selecting previously unselected package libidn12:amd64.
Preparing to unpack .../06-libidn12_1.38-4ubuntu1_amd64.deb ...
Unpacking libidn12:amd64 (1.38-4ubuntu1) ...
Selecting previously unselected package libijs-0.35:amd64.
Preparing to unpack .../07-libijs-0.35_0.35-15build2_amd64.deb ...
Unpacking libijs-0.35:amd64 (0.35-15build2) ...
Selecting previously unselected package libjbig2dec0:amd64.
Preparing to unpack .../08-libjbig2dec0_0.19-3build2_amd64.deb ...
Unpacking libjbig2dec0:amd64 (0.19-3build2) ...
Selecting previously unselected package libgs9:amd64.
Preparing to unpack .../09-libgs9 9.55.0~dfsg1-Oubuntu5.6 amd64.deb ...
Unpacking libgs9:amd64 (9.55.0~dfsg1-Oubuntu5.6) ...
Selecting previously unselected package libkpathsea6:amd64.
Preparing to unpack .../10-libkpathsea6_2021.20210626.59705-1ubuntu0.1_amd64.deb
Unpacking libkpathsea6:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package libwoff1:amd64.
Preparing to unpack .../11-libwoff1_1.0.2-1build4_amd64.deb ...
Unpacking libwoff1:amd64 (1.0.2-1build4) ...
Selecting previously unselected package dvisvgm.
Preparing to unpack .../12-dvisvgm_2.13.1-1_amd64.deb ...
Unpacking dvisvgm (2.13.1-1) ...
Selecting previously unselected package fonts-lmodern.
Preparing to unpack .../13-fonts-lmodern_2.004.5-6.1_all.deb ...
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Unpacking fonts-Imodern (2.004.5-6.1) ...
Selecting previously unselected package fonts-noto-mono.
Preparing to unpack .../14-fonts-noto-mono 20201225-1build1 all.deb ...
Unpacking fonts-noto-mono (20201225-1build1) ...
Selecting previously unselected package fonts-texgyre.
Preparing to unpack .../15-fonts-texgyre_20180621-3.1_all.deb ...
Unpacking fonts-texgyre (20180621-3.1) ...
Selecting previously unselected package libapache-pom-java.
Preparing to unpack .../16-libapache-pom-java 18-1 all.deb ...
Unpacking libapache-pom-java (18-1) ...
Selecting previously unselected package libcommons-parent-java.
Preparing to unpack .../17-libcommons-parent-java_43-1_all.deb ...
Unpacking libcommons-parent-java (43-1) ...
Selecting previously unselected package libcommons-logging-java.
Preparing to unpack .../18-libcommons-logging-java_1.2-2_all.deb ...
Unpacking libcommons-logging-java (1.2-2) ...
Selecting previously unselected package libfontenc1:amd64.
Preparing to unpack .../19-libfontenc1_1%3a1.1.4-1build3_amd64.deb ...
Unpacking libfontenc1:amd64 (1:1.1.4-1build3) ...
Selecting previously unselected package libptexenc1:amd64.
Preparing to unpack .../20-libptexenc1_2021.20210626.59705-1ubuntu0.1_amd64.deb
Unpacking libptexenc1:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package rubygems-integration.
Preparing to unpack .../21-rubygems-integration_1.18_all.deb ...
Unpacking rubygems-integration (1.18) ...
Selecting previously unselected package ruby3.0.
Preparing to unpack .../22-ruby3.0_3.0.2-7ubuntu2.4_amd64.deb ...
Unpacking ruby3.0 (3.0.2-7ubuntu2.4) ...
Selecting previously unselected package ruby-rubygems.
Preparing to unpack .../23-ruby-rubygems_3.3.5-2_all.deb ...
Unpacking ruby-rubygems (3.3.5-2) ...
Selecting previously unselected package ruby.
Preparing to unpack .../24-ruby_1%3a3.0~exp1_amd64.deb ...
Unpacking ruby (1:3.0~exp1) ...
Selecting previously unselected package rake.
Preparing to unpack .../25-rake 13.0.6-2 all.deb ...
Unpacking rake (13.0.6-2) ...
Selecting previously unselected package ruby-net-telnet.
Preparing to unpack .../26-ruby-net-telnet_0.1.1-2_all.deb ...
Unpacking ruby-net-telnet (0.1.1-2) ...
Selecting previously unselected package ruby-webrick.
Preparing to unpack .../27-ruby-webrick_1.7.0-3_all.deb ...
Unpacking ruby-webrick (1.7.0-3) ...
Selecting previously unselected package ruby-xmlrpc.
Preparing to unpack .../28-ruby-xmlrpc_0.3.2-1ubuntu0.1_all.deb ...
Unpacking ruby-xmlrpc (0.3.2-1ubuntu0.1) ...
Selecting previously unselected package libruby3.0:amd64.
```

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Preparing to unpack .../29-libruby3.0_3.0.2-7ubuntu2.4_amd64.deb ...
Unpacking libruby3.0:amd64 (3.0.2-7ubuntu2.4) ...
Selecting previously unselected package libsynctex2:amd64.
Preparing to unpack .../30-libsynctex2_2021.20210626.59705-1ubuntu0.1_amd64.deb
Unpacking libsynctex2:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package libteckit0:amd64.
Preparing to unpack .../31-libteckit0_2.5.11+ds1-1_amd64.deb ...
Unpacking libteckit0:amd64 (2.5.11+ds1-1) ...
Selecting previously unselected package libtexlua53:amd64.
Preparing to unpack .../32-libtexlua53 2021.20210626.59705-1ubuntu0.1 amd64.deb
Unpacking libtexlua53:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package libtexluajit2:amd64.
Preparing to unpack
.../33-libtexluajit2_2021.20210626.59705-1ubuntu0.1_amd64.deb ...
Unpacking libtexluajit2:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package libzzip-0-13:amd64.
Preparing to unpack .../34-libzzip-0-13_0.13.72+dfsg.1-1.1_amd64.deb ...
Unpacking libzzip-0-13:amd64 (0.13.72+dfsg.1-1.1) ...
Selecting previously unselected package xfonts-encodings.
Preparing to unpack .../35-xfonts-encodings 1%3a1.0.5-0ubuntu2 all.deb ...
Unpacking xfonts-encodings (1:1.0.5-Oubuntu2) ...
Selecting previously unselected package xfonts-utils.
Preparing to unpack .../36-xfonts-utils_1%3a7.7+6build2_amd64.deb ...
Unpacking xfonts-utils (1:7.7+6build2) ...
Selecting previously unselected package lmodern.
Preparing to unpack .../37-lmodern_2.004.5-6.1_all.deb ...
Unpacking lmodern (2.004.5-6.1) ...
Selecting previously unselected package preview-latex-style.
Preparing to unpack .../38-preview-latex-style 12.2-1ubuntu1 all.deb ...
Unpacking preview-latex-style (12.2-1ubuntu1) ...
Selecting previously unselected package tlutils.
Preparing to unpack .../39-t1utils_1.41-4build2_amd64.deb ...
Unpacking t1utils (1.41-4build2) ...
Selecting previously unselected package teckit.
Preparing to unpack .../40-teckit 2.5.11+ds1-1 amd64.deb ...
Unpacking teckit (2.5.11+ds1-1) ...
Selecting previously unselected package tex-gyre.
Preparing to unpack .../41-tex-gyre_20180621-3.1_all.deb ...
Unpacking tex-gyre (20180621-3.1) ...
Selecting previously unselected package texlive-binaries.
Preparing to unpack .../42-texlive-
binaries_2021.20210626.59705-1ubuntu0.1_amd64.deb ...
Unpacking texlive-binaries (2021.20210626.59705-1ubuntu0.1) ...
Selecting previously unselected package texlive-base.
Preparing to unpack .../43-texlive-base_2021.20220204-1_all.deb ...
Unpacking texlive-base (2021.20220204-1) ...
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Selecting previously unselected package texlive-fonts-recommended.
Preparing to unpack .../44-texlive-fonts-recommended_2021.20220204-1_all.deb ...
Unpacking texlive-fonts-recommended (2021.20220204-1) ...
Selecting previously unselected package texlive-latex-base.
Preparing to unpack .../45-texlive-latex-base 2021.20220204-1 all.deb ...
Unpacking texlive-latex-base (2021.20220204-1) ...
Selecting previously unselected package libfontbox-java.
Preparing to unpack .../46-libfontbox-java_1%3a1.8.16-2_all.deb ...
Unpacking libfontbox-java (1:1.8.16-2) ...
Selecting previously unselected package libpdfbox-java.
Preparing to unpack .../47-libpdfbox-java_1%3a1.8.16-2_all.deb ...
Unpacking libpdfbox-java (1:1.8.16-2) ...
Selecting previously unselected package texlive-latex-recommended.
Preparing to unpack .../48-texlive-latex-recommended 2021.20220204-1_all.deb ...
Unpacking texlive-latex-recommended (2021.20220204-1) ...
Selecting previously unselected package texlive-pictures.
Preparing to unpack .../49-texlive-pictures_2021.20220204-1_all.deb ...
Unpacking texlive-pictures (2021.20220204-1) ...
Selecting previously unselected package texlive-latex-extra.
Preparing to unpack .../50-texlive-latex-extra 2021.20220204-1 all.deb ...
Unpacking texlive-latex-extra (2021.20220204-1) ...
Selecting previously unselected package texlive-plain-generic.
Preparing to unpack .../51-texlive-plain-generic_2021.20220204-1_all.deb ...
Unpacking texlive-plain-generic (2021.20220204-1) ...
Selecting previously unselected package tipa.
Preparing to unpack .../52-tipa_2%3a1.3-21_all.deb ...
Unpacking tipa (2:1.3-21) ...
Selecting previously unselected package texlive-xetex.
Preparing to unpack .../53-texlive-xetex_2021.20220204-1_all.deb ...
Unpacking texlive-xetex (2021.20220204-1) ...
Setting up fonts-lato (2.0-2.1) ...
Setting up fonts-noto-mono (20201225-1build1) ...
Setting up libwoff1:amd64 (1.0.2-1build4) ...
Setting up libtexlua53:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Setting up libijs-0.35:amd64 (0.35-15build2) ...
Setting up libtexluajit2:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Setting up libfontbox-java (1:1.8.16-2) ...
Setting up rubygems-integration (1.18) ...
Setting up libzzip-0-13:amd64 (0.13.72+dfsg.1-1.1) ...
Setting up fonts-urw-base35 (20200910-1) ...
Setting up poppler-data (0.4.11-1) ...
Setting up tex-common (6.17) ...
debconf: unable to initialize frontend: Dialog
debconf: (No usable dialog-like program is installed, so the dialog based
frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line
debconf: falling back to frontend: Readline
update-language: texlive-base not installed and configured, doing nothing!
```

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Setting up libfontenc1:amd64 (1:1.1.4-1build3) ...
Setting up libjbig2dec0:amd64 (0.19-3build2) ...
Setting up libteckit0:amd64 (2.5.11+ds1-1) ...
Setting up libapache-pom-java (18-1) ...
Setting up ruby-net-telnet (0.1.1-2) ...
Setting up xfonts-encodings (1:1.0.5-Oubuntu2) ...
Setting up t1utils (1.41-4build2) ...
Setting up libidn12:amd64 (1.38-4ubuntu1) ...
Setting up fonts-texgyre (20180621-3.1) ...
Setting up libkpathsea6:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Setting up ruby-webrick (1.7.0-3) ...
Setting up fonts-lmodern (2.004.5-6.1) ...
Setting up fonts-droid-fallback (1:6.0.1r16-1.1build1) ...
Setting up ruby-xmlrpc (0.3.2-1ubuntu0.1) ...
Setting up libsynctex2:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Setting up libgs9-common (9.55.0~dfsg1-Oubuntu5.6) ...
Setting up teckit (2.5.11+ds1-1) ...
Setting up libpdfbox-java (1:1.8.16-2) ...
Setting up libgs9:amd64 (9.55.0~dfsg1-Oubuntu5.6) ...
Setting up preview-latex-style (12.2-1ubuntu1) ...
Setting up libcommons-parent-java (43-1) ...
Setting up dvisvgm (2.13.1-1) ...
Setting up libcommons-logging-java (1.2-2) ...
Setting up xfonts-utils (1:7.7+6build2) ...
Setting up libptexenc1:amd64 (2021.20210626.59705-1ubuntu0.1) ...
Setting up texlive-binaries (2021.20210626.59705-1ubuntu0.1) ...
update-alternatives: using /usr/bin/xdvi-xaw to provide /usr/bin/xdvi.bin
(xdvi.bin) in auto mode
update-alternatives: using /usr/bin/bibtex.original to provide /usr/bin/bibtex
(bibtex) in auto mode
Setting up lmodern (2.004.5-6.1) ...
Setting up texlive-base (2021.20220204-1) ...
/usr/bin/ucfr
/usr/bin/ucfr
/usr/bin/ucfr
/usr/bin/ucfr
mktexlsr: Updating /var/lib/texmf/ls-R-TEXLIVEDIST...
mktexlsr: Updating /var/lib/texmf/ls-R-TEXMFMAIN...
mktexlsr: Updating /var/lib/texmf/ls-R...
mktexlsr: Done.
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