

```
In [40]: # Put the ARN of your AWS Secrets Manager secret for your redshift cluster here:
secret_name="arn:aws:secretsmanager:us-west-1:60357419146:secret:secretkey-project-EQ0P4N"
# This will get the secret from AWS Secrets Manager.
import boto3
import json
session = boto3.session.Session()
client = session.client(
    service_name='secretsmanager'
)
get_secret_value_response = client.get_secret_value(
    SecretId=secret_name
)
if 'SecretString' in get_secret_value_response:
    connection_info = json.loads(get_secret_value_response['SecretString'])
else:
    print("ERROR: no secret data found")
```

```
In [41]: expected_keys = set(['dbname','host','dbClusterIdentifier','password', 'port', 'engine', 'username'])
if not expected_keys.issubset(connection_info.keys()):
    print("Expected values for ",expected_keys)
    print("Received values for ",set(connection_info.keys()))
    print("Please adjust query or assignment as required!")
```

```
In [42]: import time
import psycopg2
con=psycopg2.connect(
    dbname = connection_info["dbname"],
    host = connection_info["host"],
    port = connection_info["port"],
    user = connection_info["username"],
    password = connection_info["password"]
)
```

```
In [43]: cur = con.cursor()
cur.execute("select * from usa_housing_population limit 5")
res = cur.fetchall()
print(res)
cur.close()
```

```
{{'cleveland', 'OH', '11/1/17', 142952, 44102, 191450, 0.161, 1.7389, 133, -0.0764, 0.031, None, 0.163265306, 0.17931
0.945, 24, -0.25, -0.1429, 4, None, None, 20, -0.375, -0.1667, 65, 0.1917, 3.3333, 115.2431818, 0.1815, 1.5241, 1628,
-0.0317, 0.1922, 284958.5734, 0.0704, 1.0099, 199, -0.0246, 0.375, 0.488721805}, ('greenville', 'Nc', '11/1/17', 13224
1, 27834, 142409, -0.0659, -0.0659, 126, -0.0455, -0.2921, 88, -0.257142857, -0.293172691, 24, -0.1429, -0.25, 4, None,
None, 29, 0.6667, -0.375, 90, -0.0816, 0.1688, 86.73016101, -0.0537, 0.0581, 1604, -0.0795, 0.0409, 203423.6565, -0.0
572, 0.0906, 218, -0.0609, -0.1529, 0.714285714}, ('tampa', 'FL', '11/1/17', 201871, 33616, 259900, 0.0, 0.0554, 73,
-0.0135, -0.0519, 69, -0.066092715, 0.568181818, 28, 0.1667, 0.0, 4, 0.0, None, 20, -0.5455, -0.1667, 45, -0.1346, 0.
25, 172.6422387, -0.0024, 0.0661, None, -0.0245, 0.058, 323986.6429, 0.0136, 0.0115, 118, -0.0635, 0.0442, 0.61643835
6}, ('miami', 'FL', '11/1/17', 278983, 33185, 439500, 0.0163, 0.0222, 84, 0.0244, -0.1429, 62, -0.107913669, -0.21019
1083, 28, -0.125, -0.3, 4, 0.0, 0.0, 20, 0.25, 0.6667, 61, -0.0615, 0.0, 189.8023434, 0.0189, 0.0895, None, 0.0905, -0.
0604, 432023.0409, 0.0214, 0.0107, 145, -0.0136, -0.0881, 0.726190476}, ('denver', 'CO', '11/1/17', 408911, 80231,
427250, 0.5399, -0.1453, 48, -0.25, -0.0843, 36, 0.2, -0.152041176, 32, -0.3846, 0.1429, 4, None, None, 20, -0.2857,
0.0, 1, -0.5, 0.0, 172.9302533, 0.0053, 0.0435, 2642, 0.5783, -0.0833, 618434.2169, 0.1667, -0.1487, 49, -0.2576, -0.
0926, 0.0208333333})
```

```
In [44]: import pandas as pd
import pandas.io.sql as sqlio
import psycopg2
import pandas as pd
import numpy as np
import seaborn as sns
import datetime
import matplotlib.pyplot as plt
%matplotlib inline
```

```
In [45]: sql = "select * from usa_housing_population;"
housing_population= sqlio.read_sql_query(sql, con)
```

```
In [46]: housing_population
Out[46]:
```

	city	state	date	population	postal_code	median_listing_price	median_listing_price_mm	median_listing_price_yy	active_listing_count	active_l
0	cleveland	OH	11/1/17	142952	44102	191450.0	0.1610	1.7389	133	
1	greenville	NC	11/1/17	132241	27834	142400.0	-0.0659	0.1005	126	
2	tampa	FL	11/1/17	201871	33616	259900.0	0.0000	0.0554	73	
3	miami	FL	11/1/17	278083	33185	439500.0	0.0163	0.0222	84	
4	denver	CO	11/1/17	408911	80231	427250.0	0.5399	-0.1453	48	
...
214084	fort myers	FL	7/1/21	305212	33913	489900.0	0.1134	NaN	63	
214085	new york	NY	7/1/21	554085	10065	1925000.0	0.0405	0.0280	443	
214086	sioux falls	SD	7/1/21	261064	57106	265000.0	0.0004	NaN	190	
214087	cullman	AL	7/1/21	116254	35068	229000.0	-0.1820	0.1456	27	
214088	muncie	IN	7/1/21	156519	47303	155000.0	0.3784	NaN	41	

214089 rows × 39 columns

```
In [47]: housing_population.describe()
Out[47]:
```

	population	postal_code	median_listing_price	median_listing_price_mm	median_listing_price_yy	active_listing_count	active_listing_count_mm	acti
count	2.140890e+05	214089.000000	1.989090e+05	171010.000000	156691.000000	214089.000000	171010.000000	
mean	2.254718e+05	54865.608761	3.262474e+05	0.008634	0.069869	91.396980	0.002834	
std	1.545481e+05	26330.251459	3.530873e+05	0.102221	0.170113	91.231932	0.232148	
min	3.866300e+04	1104.000000	2.500000e+03	-0.960900	-0.929200	13.000000	-0.727300	
25%	1.457490e+05	32966.000000	1.500000e+05	-0.027600	-0.020800	36.000000	-0.085500	
50%	1.967090e+05	52246.000000	2.390000e+05	0.000000	0.055600	65.000000	-0.009600	
75%	2.670410e+05	78260.000000	3.750000e+05	0.035800	0.144300	115.000000	0.068700	
max	1.455985e+06	99950.000000	6.622500e+06	14.018800	7.410900	1536.000000	68.000000	

8 rows × 36 columns

```
In [48]: housing_population= housing_population.sort_values(['state', 'date'], ascending = (True, True)).reset_index(drop = T
True)
```

```
In [49]: housing_population['median_square_feet_yy'].mean()
housing_population['average_listing_price'].mean()
Out[49]: 422409.51317344076
```

```
In [50]: housing_population[['average_listing_price','median_square_feet_yy','median_listing_price_mm','median_listing_p
rice_yy']] = housing_population[['average_listing_price','median_square_feet_yy','total_listing_count_mm','total_lis
ting_count_yy']].fillna(value=housing_population[['average_listing_price','median_square_feet_yy','median_listing_p
rice_mm','median_listing_price_yy']].mean())
```

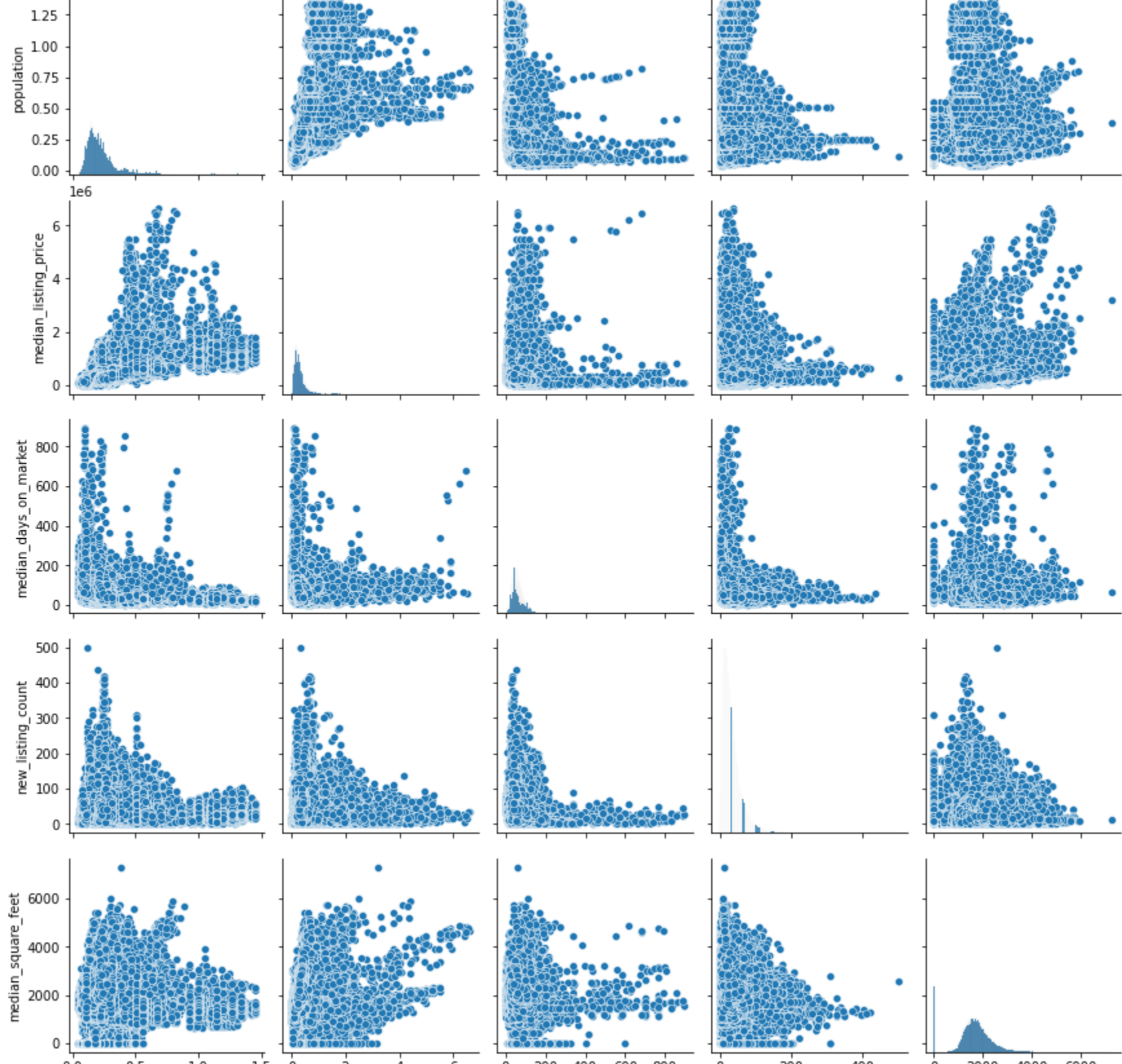
```
In [51]: housing_population[['average_listing_price_mm','average_listing_price_yy','total_listing_count_mm','total_listing_co
unt_yy']] = housing_population[['average_listing_price_mm','average_listing_price_yy','total_listing_count_mm','total_l
isting_count_yy']].fillna(value=housing_population[['average_listing_price_mm','average_listing_price_yy','total_l
isting_count_mm','total_listing_count_yy']].mean())
```

```
In [52]: filters = [
    (housing_population['average_listing_price_yy'] < 0) & (housing_population['median_listing_price_yy'] < 0),
    (housing_population['average_listing_price_mm'] < 0) & (housing_population['median_listing_price_mm'] < 0),
    ]
values = ["Down from last year", "Down from last month"]
housing_population["category"] = np.select(filters, values, default="Increase in prices" )
print('Price ratios for all the records:')
housing_population['category'].value_counts()/len(housing_population['category'])*100
#compared to previous years, there has been an increase in prices by 65.36%
Price ratios for all the records:
```

```
Out[52]: Increase in prices    65.388693
Down from last month    19.763276
Down from last year    14.848030
Name: category, dtype: float64
```

```
In [54]: housing_population[['month', 'day', 'year']] = housing_population["date"].str.split("/", expand = True)
```

```
In [55]: data1=housing_population[['population', 'median_listing_price', 'median_days_on_market', 'new_listing_count', 'median_s
quare_feet']]
sns.pairplot(data1.iloc[:, :5])
plt.show()
```



```
In [56]: housing_population.iloc[:, :].hist(figsize=(25,35))
plt.show()
```

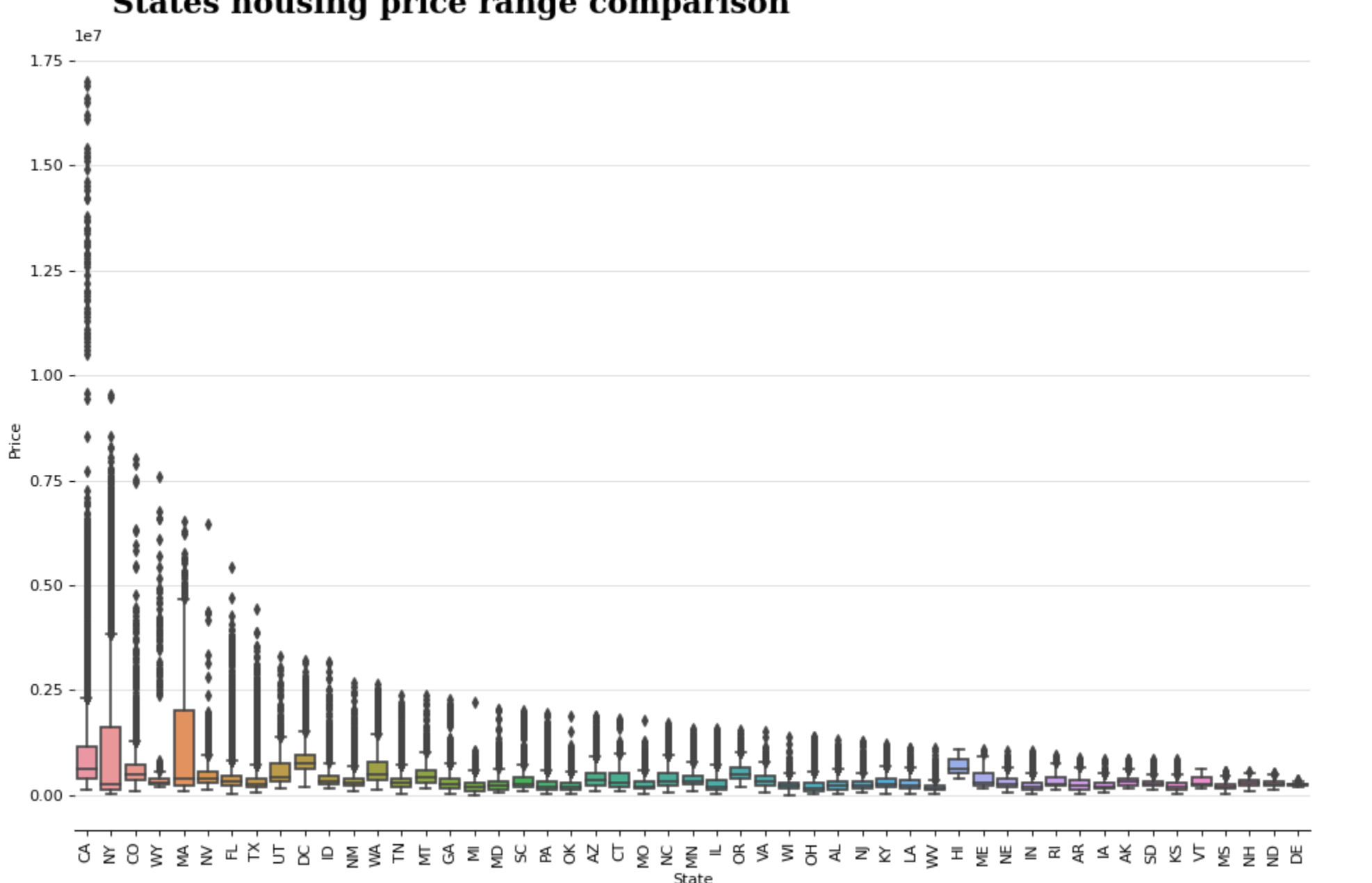


```
In [57]: fig, ax = plt.subplots(1,1, figsize=(12, 8), dpi=80)
fig.text(0.09, 1, 'States housing price range comparison', fontsize=20, fontweight='bold', fontfamily='serif')

#plt.figure(figsize=(10, 8), dpi=80)
box_plot = sns.boxplot(x = 'state', y = 'average_listing_price', data = housing_population.sort_values('average_listin
g_price', ascending=False))
plt.ylabel('Price')
plt.xlabel('State')

ax = box_plot.axes
lines = ax.get_lines()
categories = ax.get_xticks()
ax.tick_params(axis='x', rotation=90)

for s in ['top', 'left', 'right']:
    ax.spines[s].set_visible(False)
ax.grid(axis='y', linestyle='-', alpha=0.4)
box_plot.figure.tight_layout()
#fig = box_plot.get_figure()
```



```
In [59]: df_reduced = housing_population[['average_listing_price', 'median_listing_price', 'median_listing_price_per_square_fo
ot', 'price_reduced_count', 'active_listing_count', 'new_listing_count', 'median_days_on_market', 'year']]
sns.heatmap(df_reduced.corr(), annot = True)
```

```
Out[59]: <AxesSubplot: >
```

	average_listing_price	median_listing_price	median_listing_price_per_square_foot	price_reduced_count	active_listing_count	new_listing_count	median_days_on_market
average_listing_price	1	0.94	0.78	0.061	0.14	0.1	0.021
median_listing_price	0.94	1	0.85	0.048	0.098	0.1	0.029
median_listing_price_per_square_foot	0.78	0.85	1	0.00046	0.052	0.063	0.061
price_reduced_count	0.061	0.048	0.00046	1	0.79	0.76	0.017
active_listing_count	0.14	0.098	0.052	0.79	1	0.72	0.21
new_listing_count	0.1	0.1	0.063	0.76	0.72	1	0.18
median_days_on_market	0.021	0.029	0.061	0.017	0.21	0.18	1

```
In [ ]:
```