

```
data["price"].max()
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-1-8fe122fbccf5> in <module>()
----> 1 data["price"].max()

NameError: name 'data' is not defined
```

SEARCH STACK OVERFLOW

```
import pandas as pd
data = pd.read_csv("http://hackveda.in/sistec/Housing_Modified.csv")
print(data)
```

| | price | lotsize | bedrooms | bathrms | ... | gashw | airco | garagepl | prefarea |
|-----|----------|---------|----------|---------|-----|-------|-------|----------|----------|
| 0 | 42000.0 | 5850 | 3 | 1 | ... | no | no | 1 | no |
| 1 | 38500.0 | 4000 | 2 | 1 | ... | no | no | 0 | no |
| 2 | 49500.0 | 3060 | 3 | 1 | ... | no | no | 0 | no |
| 3 | 60500.0 | 6650 | 3 | 1 | ... | no | no | 0 | no |
| 4 | 61000.0 | 6360 | 2 | 1 | ... | no | no | 0 | no |
| .. | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 541 | 91500.0 | 4800 | 3 | 2 | ... | no | yes | 0 | no |
| 542 | 94000.0 | 6000 | 3 | 2 | ... | no | yes | 0 | no |
| 543 | 103000.0 | 6000 | 3 | 2 | ... | no | yes | 1 | no |
| 544 | 105000.0 | 6000 | 3 | 2 | ... | no | yes | 1 | no |
| 545 | 105000.0 | 6000 | 3 | 1 | ... | no | yes | 1 | no |

[546 rows x 12 columns]

```
data["price"].max()
data["price"].min()
data["price"].mean()
data["price"].sum()
```

37194392.0

```
data["bedrooms"].unique()
data["airco"].unique()

array(['no', 'yes'], dtype=object)
```

```
data.line("price","lotsize")
```

```

-----
AttributeError                                Traceback (most recent call last)
<ipython-input-13-80e5c431fdb5> in <module>()
----> 1 data.line("price","lotsize")

/usr/local/lib/python3.6/dist-packages/pandas/core/generic.py in __getattr__(self, name)
    5272         if self._info_axis._can_hold_identifiers_and_holds_name(name):
    5273             return self[name]
-> 5274         return object.__getattr__(self, name)
    5275

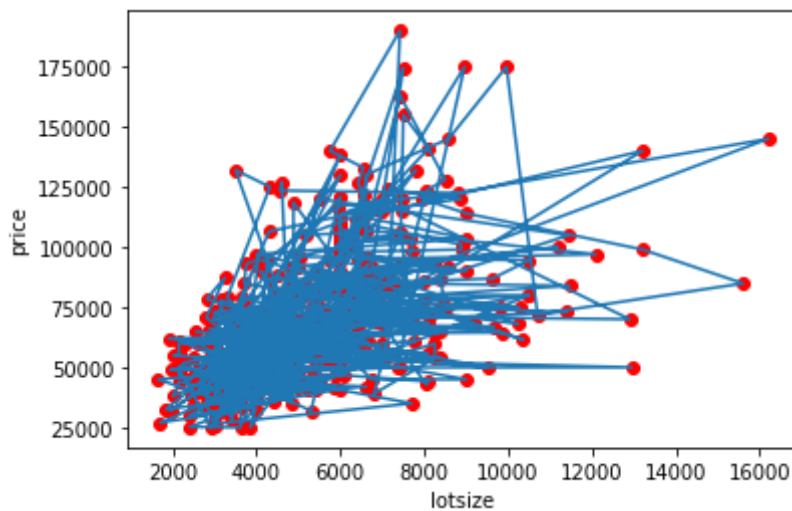
```

```

import matplotlib.pyplot as plt
plt.scatter(data["lotsize"],data["price"],color="red")
plt.plot(data["lotsize"],data["price"])
plt.xlabel("lotsize")
plt.ylabel("price")

```

Text(0, 0.5, 'price')



```
data.head(3)
```

| | price | lotsize | bedrooms | bathrms | stories | driveway | recroom | fullbase | gashw |
|---|---------|---------|----------|---------|---------|----------|---------|----------|-------|
| 0 | 42000.0 | 5850 | 3 | 1 | two | yes | no | yes | no |
| 1 | 38500.0 | 4000 | 2 | 1 | one | yes | no | no | no |
| 2 | 49500.0 | 3060 | 3 | 1 | one | yes | no | no | no |

```

#convert text to numbers using Label Binarizer
import sklearn.preprocessing as pp
lb = pp.LabelBinarizer()
data.driveway = lb.fit_transform(data.driveway)
data.recroom = lb.fit_transform(data.recroom)
data.head(15)

```

| | price | lotsize | bedrooms | bathrms | stories | driveway | recroom | fullbase | gashw |
|----|---------|---------|----------|---------|---------|----------|---------|----------|-------|
| 0 | 42000.0 | 5850 | 3 | 1 | two | 1 | 0 | yes | no |
| 1 | 38500.0 | 4000 | 2 | 1 | one | 1 | 0 | no | no |
| 2 | 49500.0 | 3060 | 3 | 1 | one | 1 | 0 | no | no |
| 3 | 60500.0 | 6650 | 3 | 1 | two | 1 | 1 | no | no |
| 4 | 61000.0 | 6360 | 2 | 1 | one | 1 | 0 | no | no |
| 5 | 66000.0 | 4160 | 3 | 1 | one | 1 | 1 | yes | no |
| 6 | 66000.0 | 3880 | 3 | 2 | two | 1 | 0 | yes | no |
| 7 | 69000.0 | 4160 | 3 | 1 | three | 1 | 0 | no | no |
| 8 | 83800.0 | 4800 | 3 | 1 | one | 1 | 1 | yes | no |
| 9 | 88500.0 | 5500 | 3 | 2 | four | 1 | 1 | no | no |
| 10 | 90000.0 | 7200 | 3 | 2 | one | 1 | 0 | yes | no |

```
data.corr()
```

| | price | lotsize | bedrooms | bathrms | driveway | recroom | garagepl |
|----------|----------|----------|-----------|----------|-----------|----------|----------|
| price | 1.000000 | 0.535796 | 0.366447 | 0.516719 | 0.297167 | 0.254960 | 0.383302 |
| lotsize | 0.535796 | 1.000000 | 0.151851 | 0.193833 | 0.288778 | 0.140327 | 0.352872 |
| bedrooms | 0.366447 | 0.151851 | 1.000000 | 0.373769 | -0.011996 | 0.080492 | 0.139117 |
| bathrms | 0.516719 | 0.193833 | 0.373769 | 1.000000 | 0.041955 | 0.126892 | 0.178178 |
| driveway | 0.297167 | 0.288778 | -0.011996 | 0.041955 | 1.000000 | 0.091959 | 0.203682 |
| recroom | 0.254960 | 0.140327 | 0.080492 | 0.126892 | 0.091959 | 1.000000 | 0.038122 |
| garagepl | 0.383302 | 0.352872 | 0.139117 | 0.178178 | 0.203682 | 0.038122 | 1.000000 |

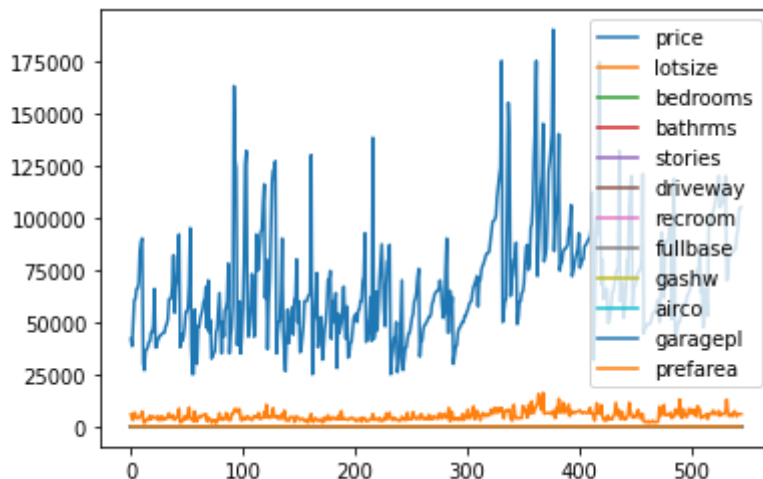
```
data = pd.read_csv("Housing_Modified_prepared.csv")
data.corr()
```

| | price | lotsize | bedrooms | bathrms | stories | driveway | recroom | full |
|-----------------|----------|----------|----------|----------|----------|----------|----------|-------|
| price | 1.000000 | 0.535796 | 0.366447 | 0.516719 | 0.421190 | 0.297167 | 0.254960 | 0.180 |
| lotsize | 0.535796 | 1.000000 | 0.151851 | 0.193833 | 0.083675 | 0.288778 | 0.140327 | 0.04 |
| bedrooms | 0.366447 | 0.151851 | 1.000000 | 0.070760 | 0.407074 | 0.011006 | 0.000400 | 0.00 |

```
pp.LabelEncoder()
```

```
import pandas as pd
data = pd.read_csv("Housing_Modified_prepared.csv")
data.plot()
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f67c8866da0>



```
#Data transformation using normalization
#Implement standardization
# Formula : (X -Xmean)/Xstd
# X = Independent variable
# Xmean = Mean of X variable
# Xstd = Standard Deviation of X independent variable
print("Columns in data : ", data.columns)
print("Max Value of lotsize before transformation :", data["lotsize"].max())
print("Min Value of lotsize before transformation :", data["lotsize"].min())
print("Mean Value of lotsize before transformation :", data["lotsize"].mean())
```

```
#Perform transformation using standard score/Standardization
X = data["lotsize"] #Select the independent variable
Xmean = X.mean() # calculate the mean of independent variable
Xstd = X.std() # calculate the standard deviation of independent variable
Xnorm = (X - Xmean) / Xstd
print("X normalized after transformation\n")
Xnorm
```

```
print("Min value of Xnorm", Xnorm.min())
print("Max value of Xnorm", Xnorm.max())
```

```
Columns in data : Index(['price', 'lotsize', 'bedrooms', 'bathrms', 'stories', 'driveway', 'recroom', 'fullbase', 'gashw', 'airco', 'garagepl', 'prefarea'],
dtype='object')
Max Value of lotsize before transformation : 16200
```

```
Min Value of lotsize before transformation : 1650
Mean Value of lotsize before transformation : 5150.2655677655675
X normalized after transformation
```

```
Min value of Xnorm -1.6143954439482222
Max value of Xnorm 5.096367855203785
```

```
independent_variables = data.columns
independent_variables = independent_variables.delete(0)
print("Independent Variables : ", independent_variables)
```

```
Independent Variables : Index(['lotsize', 'bedrooms', 'bathrms', 'stories', 'driveway',
                             'fullbase', 'gashw', 'airco', 'garagepl', 'prefarea'],
                             dtype='object')
```

```
data[independent_variables]
```

| | lotsize | bedrooms | bathrms | stories | driveway | recroom | fullbase | gashw | airco |
|------------|---------|----------|---------|---------|----------|---------|----------|-------|-------|
| 0 | 5850 | 3 | 1 | 2 | 1 | 0 | 1 | 0 | 0 |
| 1 | 4000 | 2 | 1 | 1 | 1 | 0 | 0 | 0 | 0 |
| 2 | 3060 | 3 | 1 | 1 | 1 | 0 | 0 | 0 | 0 |
| 3 | 6650 | 3 | 1 | 2 | 1 | 1 | 0 | 0 | 0 |
| 4 | 6360 | 2 | 1 | 1 | 1 | 0 | 0 | 0 | 0 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 541 | 4800 | 3 | 2 | 4 | 1 | 1 | 0 | 0 | 1 |
| 542 | 6000 | 3 | 2 | 4 | 1 | 0 | 0 | 0 | 1 |
| 543 | 6000 | 3 | 2 | 4 | 1 | 1 | 0 | 0 | 1 |
| 544 | 6000 | 3 | 2 | 2 | 1 | 1 | 0 | 0 | 1 |
| 545 | 6000 | 3 | 1 | 2 | 1 | 0 | 0 | 0 | 1 |

546 rows × 11 columns

```
X = data[independent_variables] # Independent variables
Y = data["price"] # Dependent variable
X
Y
```

```
# perform standardization on independent variables
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
Xnorm = scale.fit_transform(X)
print("Normalized data using standardization")
Xnorm
```

Normalized data using standardization

```
array([[ 0.32302806,  0.0472349 , -0.5694948 , ..., -0.68103375,
         0.35756661, -0.55337157],
       [-0.53101296, -1.31014696, -0.5694948 , ..., -0.68103375,
        -0.80452487, -0.55337157],
       [-0.96495812,  0.0472349 , -0.5694948 , ..., -0.68103375,
        -0.80452487, -0.55337157],
       ...,
       [ 0.39227462,  0.0472349 ,  1.42373699, ...,  1.46835601,
         0.35756661, -0.55337157],
       [ 0.39227462,  0.0472349 ,  1.42373699, ...,  1.46835601,
         0.35756661, -0.55337157],
       [ 0.39227462,  0.0472349 , -0.5694948 , ...,  1.46835601,
         0.35756661, -0.55337157]])
```

```
# feature scaling to normalize data on a common scale (0-1)
# feature scaling is also known as Min-Max Scaling
# formula for min-max scaling is :  $X_{norm} = (X - X_{min}) / (X_{max} - X_{min})$ 
print("Max Value of lotsize :", data["lotsize"].max())
print("Min Value of lotsize :", data["lotsize"].min())
data["lotsize"].head(2)
```

```
Max Value of lotsize : 16200
Min Value of lotsize : 1650
0    5850
1    4000
Name: lotsize, dtype: int64
```

```
X = data["lotsize"] # select the variable
Xmin = X.min() #calculate the min value of data variable
Xmax = X.max() #calculate the max value of data variable
Xnorm = (X - Xmin)/(Xmax - Xmin)
print("Normalized values after feature scaling\n")
Xnorm
```

```
print("Min value of Xnorm", Xnorm.min())
print("Max value of Xnorm", Xnorm.max())
```

Normalized values after feature scaling

```
Min value of Xnorm 0.0
Max value of Xnorm 1.0
```

```
X = data[independent_variables]
# feature scaling using Min-Max Scaling
Xmin = X.min() #calculate the min value of data variable
Xmax = X.max() #calculate the max value of data variable
Xnorm = (X - Xmin)/(Xmax - Xmin)
print("Feature scaled independent variables after transformations\n")
Xnorm
```

Feature scaled independent variables after transformations

| | lotsize | bedrooms | bathrms | stories | driveway | recroom | fullbase | gashw | airco |
|------------|----------|----------|----------|----------|----------|---------|----------|-------|-------|
| 0 | 0.288660 | 0.4 | 0.000000 | 0.333333 | 1.0 | 0.0 | 1.0 | 0.0 | 0.0 |
| 1 | 0.161512 | 0.2 | 0.000000 | 0.000000 | 1.0 | 0.0 | 0.0 | 0.0 | 0.0 |
| 2 | 0.096907 | 0.4 | 0.000000 | 0.000000 | 1.0 | 0.0 | 0.0 | 0.0 | 0.0 |
| 3 | 0.343643 | 0.4 | 0.000000 | 0.333333 | 1.0 | 1.0 | 0.0 | 0.0 | 0.0 |
| 4 | 0.323711 | 0.2 | 0.000000 | 0.000000 | 1.0 | 0.0 | 0.0 | 0.0 | 0.0 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 541 | 0.216495 | 0.4 | 0.333333 | 1.000000 | 1.0 | 1.0 | 0.0 | 0.0 | 1.0 |
| 542 | 0.298969 | 0.4 | 0.333333 | 1.000000 | 1.0 | 0.0 | 0.0 | 0.0 | 1.0 |
| 543 | 0.298969 | 0.4 | 0.333333 | 1.000000 | 1.0 | 1.0 | 0.0 | 0.0 | 1.0 |
| 544 | 0.298969 | 0.4 | 0.333333 | 0.333333 | 1.0 | 1.0 | 0.0 | 0.0 | 1.0 |
| 545 | 0.298969 | 0.4 | 0.000000 | 0.333333 | 1.0 | 0.0 | 0.0 | 0.0 | 1.0 |

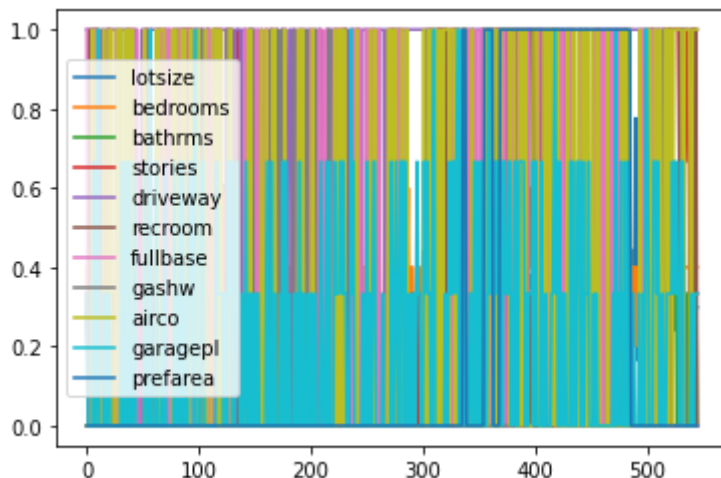
546 rows x 11 columns

```
X = data[independent_variables]
# feature scaling using Min-Max Scaling
Xmin = X.min() #calculate the min value of data variable
Xmax = X.max() #calculate the max value of data variable
Xnorm = (X - Xmin)/(Xmax - Xmin)
print("Feature scaled independent variables after transformations\n")

# Make a plot of normalized values
Xnorm.plot()
```

Feature scaled independent variables after transformations

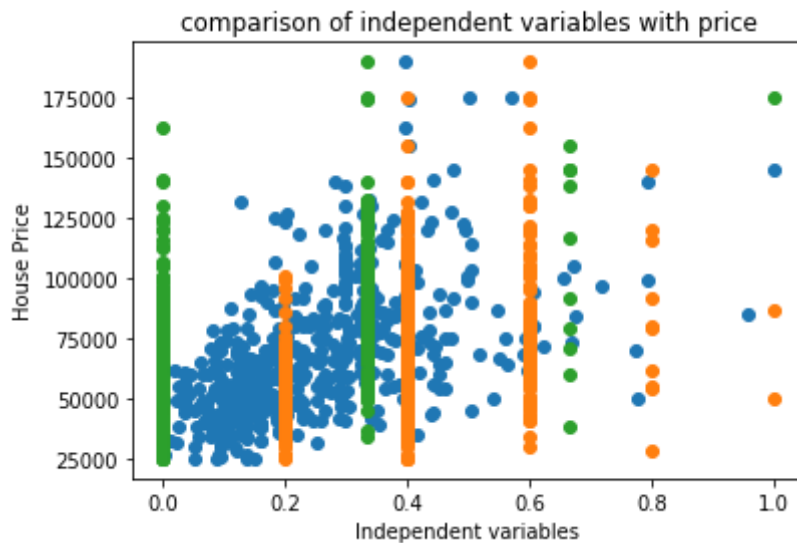
<matplotlib.axes._subplots.AxesSubplot at 0x7fd2b0c00a20>



```
# Compare independent variables and their influence on price
import matplotlib.pyplot as plt
plt.scatter(Xnorm["lotsize"], data["price"])
plt.scatter(Xnorm["bedrooms"], data["price"])
plt.scatter(Xnorm["bathrms"], data["price"])
plt.scatter(Xnorm["stories"], data["price"])
plt.scatter(Xnorm["driveway"], data["price"])
plt.scatter(Xnorm["recroom"], data["price"])
plt.scatter(Xnorm["fullbase"], data["price"])
plt.scatter(Xnorm["gashw"], data["price"])
plt.scatter(Xnorm["airco"], data["price"])
plt.scatter(Xnorm["garagepl"], data["price"])
plt.scatter(Xnorm["prefarea"], data["price"])
```

```
plt.scatter(xnorm["bathrms"], data["price"])
plt.title("comparison of independent variables with price")
plt.xlabel("Independent variables")
plt.ylabel("House Price")
```

```
Text(0, 0.5, 'House Price')
```



```
# perform feature scaling using min-max scaler
from sklearn.preprocessing import MinMaxScaler
scale = MinMaxScaler()
Xnorm = scale.fit_transform(X)
print("transformation using min-max scaler\n")
Xnorm
```

```
transformation using min-max scaler
```

```
array([[0.28865979, 0.4      , 0.      , ..., 0.      , 0.33333333,
        0.      ],
       [0.16151203, 0.2      , 0.      , ..., 0.      , 0.      ,
        0.      ],
       [0.09690722, 0.4      , 0.      , ..., 0.      , 0.      ,
        0.      ],
       ...,
       [0.29896907, 0.4      , 0.33333333, ..., 1.      , 0.33333333,
        0.      ],
       [0.29896907, 0.4      , 0.33333333, ..., 1.      , 0.33333333,
        0.      ],
       [0.29896907, 0.4      , 0.      , ..., 1.      , 0.33333333,
        0.      ]])
```

```
# Create a correlation matrix using Seaborn and Matplotlib
import seaborn as sns
```

```
# Set up the matplotlib figure
size = max(10, len(data.columns)/2.)
f, ax = plt.subplots(figsize=(size, size))
```

```
# Draw the heatmap with the mask and correct aspect ratio
sns.heatmap(data.corr(), annot=True, square=True, linewidths=.5, cbar_kws={"shrink": 0.5},
```



<matplotlib.axes._subplots.AxesSubplot at 0x7fd2af7dce48>

