

Shellya Nur Atqiya 1903685

IMPORT DATA

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np

pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)

train = pd.read_csv('/content/train.csv')
test = pd.read_csv('/content/test.csv')

#show top 5 rows
train.head(5)
```

	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape
0	1	60	RL	65.0	8450	Pave	NaN	Reg
1	2	20	RL	80.0	9600	Pave	NaN	Reg
2	3	60	RL	68.0	11250	Pave	NaN	IR1
3	4	70	RL	60.0	9550	Pave	NaN	IR1
4	5	60	RL	84.0	14260	Pave	NaN	IR1

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1
0	Lvl	AllPub	Inside	Gtl	CollgCr	Norm
1	Lvl	AllPub	FR2	Gtl	Veenker	Feedr
2	Lvl	AllPub	Inside	Gtl	CollgCr	Norm
3	Lvl	AllPub	Corner	Gtl	Crawfor	Norm
4	Lvl	AllPub	FR2	Gtl	NoRidge	Norm

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt
0	Norm	1Fam	2Story	7	5	2003
1	Norm	1Fam	1Story	6	8	1976
2	Norm	1Fam	2Story	7	5	2001
3	Norm	1Fam	2Story	7	5	1915

4	Norm	1Fam	2Story	8	5	2000
---	------	------	--------	---	---	------

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType
\						
0	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace
1	1976	Gable	CompShg	MetalSd	MetalSd	None
2	2002	Gable	CompShg	VinylSd	VinylSd	BrkFace
3	1970	Gable	CompShg	Wd Sdng	Wd Shng	None
4	2000	Gable	CompShg	VinylSd	VinylSd	BrkFace

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond
BsmtExposure \						
0	196.0	Gd	TA	PConc	Gd	TA
No						
1	0.0	TA	TA	CBlock	Gd	TA
Gd						
2	162.0	Gd	TA	PConc	Gd	TA
Mn						
3	0.0	TA	TA	BrkTil	TA	Gd
No						
4	350.0	Gd	TA	PConc	Gd	TA
Av						

	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	BsmtUnfSF
TotalBsmtSF \					
0	GLQ	706	Unf	0	150
856					
1	ALQ	978	Unf	0	284
1262					
2	GLQ	486	Unf	0	434
920					
3	ALQ	216	Unf	0	540
756					
4	GLQ	655	Unf	0	490
1145					

	Heating	HeatingQC	CentralAir	Electrical	1stFlrSF	2ndFlrSF
LowQualFinSF \						
0	GasA	Ex	Y	SBrkr	856	854
0						
1	GasA	Ex	Y	SBrkr	1262	0
0						

2	GasA	Ex	Y	SBrkr	920	866
0						
3	GasA	Gd	Y	SBrkr	961	756
0						
4	GasA	Ex	Y	SBrkr	1145	1053
0						

	GrLivArea	BsmtFullBath	BsmtHalfBath	FullBath	HalfBath
BedroomAbvGr	\				
0	1710	1	0	2	1
3					
1	1262	0	1	2	0
3					
2	1786	1	0	2	1
3					
3	1717	1	0	1	0
3					
4	2198	1	0	2	1
4					

	KitchenAbvGr	KitchenQual	TotRmsAbvGrd	Functional	Fireplaces
FireplaceQu	\				
0	1	Gd	8	Typ	0
NaN					
1	1	TA	6	Typ	1
TA					
2	1	Gd	6	Typ	1
TA					
3	1	Gd	7	Typ	1
Gd					
4	1	Gd	9	Typ	1
TA					

	GarageType	GarageYrBlt	GarageFinish	GarageCars	GarageArea
GarageQual	\				
0	Attchd	2003.0	RFn	2	548
TA					
1	Attchd	1976.0	RFn	2	460
TA					
2	Attchd	2001.0	RFn	2	608
TA					
3	Detchd	1998.0	Unf	3	642
TA					
4	Attchd	2000.0	RFn	3	836
TA					

	GarageCond	PavedDrive	WoodDeckSF	OpenPorchSF	EnclosedPorch
3SsnPorch	\				
0	TA	Y	0	61	0
0					

1	TA	Y	298	0	0
0					
2	TA	Y	0	42	0
0					
3	TA	Y	0	35	272
0					
4	TA	Y	192	84	0
0					

	ScreenPorch	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold
YrSold \							
0	0	0	NaN	NaN	NaN	0	2
2008							
1	0	0	NaN	NaN	NaN	0	5
2007							
2	0	0	NaN	NaN	NaN	0	9
2008							
3	0	0	NaN	NaN	NaN	0	2
2006							
4	0	0	NaN	NaN	NaN	0	12
2008							

	SaleType	SaleCondition	SalePrice
0	WD	Normal	208500
1	WD	Normal	181500
2	WD	Normal	223500
3	WD	Abnorml	140000
4	WD	Normal	250000

```
#drop column "Id"
```

```
train = train.drop(["Id"], axis=1)
```

```
#count rows and columns
```

```
train.shape
```

```
(1460, 80)
```

```
#show columns
```

```
train.columns
```

```
Index(['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
       'Alley',
       'LotShape', 'LandContour', 'Utilities', 'LotConfig',
       'LandSlope',
       'Neighborhood', 'Condition1', 'Condition2', 'BldgType',
       'HouseStyle',
       'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd',
       'RoofStyle',
       'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType',
       'MasVnrArea',
       'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', 'BsmtCond',
```

```

        'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2',
        'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating',
'HeatingQC',
        'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF',
'LowQualFinSF',
        'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath',
'HalfBath',
        'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd',
        'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType',
'GarageYrBlt',
        'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual',
'GarageCond',
        'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch',
'3SsnPorch',
        'ScreenPorch', 'PoolArea', 'PoolQC', 'Fence', 'MiscFeature',
'MiscVal',
        'MoSold', 'YrSold', 'SaleType', 'SaleCondition', 'SalePrice'],
dtype='object')

```

Handling Missing Values

```
train.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 80 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   MSSubClass            1460 non-null  int64
 1   MSZoning              1460 non-null  object
 2   LotFrontage          1201 non-null  float64
 3   LotArea              1460 non-null  int64
 4   Street               1460 non-null  object
 5   Alley               91 non-null    object
 6   LotShape             1460 non-null  object
 7   LandContour          1460 non-null  object
 8   Utilities            1460 non-null  object
 9   LotConfig            1460 non-null  object
10   LandSlope            1460 non-null  object
11   Neighborhood         1460 non-null  object
12   Condition1           1460 non-null  object
13   Condition2           1460 non-null  object
14   BldgType             1460 non-null  object
15   HouseStyle           1460 non-null  object
16   OverallQual          1460 non-null  int64
17   OverallCond          1460 non-null  int64
18   YearBuilt            1460 non-null  int64
19   YearRemodAdd         1460 non-null  int64
20   RoofStyle            1460 non-null  object
21   RoofMatl            1460 non-null  object
22   Exterior1st         1460 non-null  object

```

23	Exterior2nd	1460	non-null	object
24	MasVnrType	1452	non-null	object
25	MasVnrArea	1452	non-null	float64
26	ExterQual	1460	non-null	object
27	ExterCond	1460	non-null	object
28	Foundation	1460	non-null	object
29	BsmtQual	1423	non-null	object
30	BsmtCond	1423	non-null	object
31	BsmtExposure	1422	non-null	object
32	BsmtFinType1	1423	non-null	object
33	BsmtFinSF1	1460	non-null	int64
34	BsmtFinType2	1422	non-null	object
35	BsmtFinSF2	1460	non-null	int64
36	BsmtUnfSF	1460	non-null	int64
37	TotalBsmtSF	1460	non-null	int64
38	Heating	1460	non-null	object
39	HeatingQC	1460	non-null	object
40	CentralAir	1460	non-null	object
41	Electrical	1459	non-null	object
42	1stFlrSF	1460	non-null	int64
43	2ndFlrSF	1460	non-null	int64
44	LowQualFinSF	1460	non-null	int64
45	GrLivArea	1460	non-null	int64
46	BsmtFullBath	1460	non-null	int64
47	BsmtHalfBath	1460	non-null	int64
48	FullBath	1460	non-null	int64
49	HalfBath	1460	non-null	int64
50	BedroomAbvGr	1460	non-null	int64
51	KitchenAbvGr	1460	non-null	int64
52	KitchenQual	1460	non-null	object
53	TotRmsAbvGrd	1460	non-null	int64
54	Functional	1460	non-null	object
55	Fireplaces	1460	non-null	int64
56	FireplaceQu	770	non-null	object
57	GarageType	1379	non-null	object
58	GarageYrBlt	1379	non-null	float64
59	GarageFinish	1379	non-null	object
60	GarageCars	1460	non-null	int64
61	GarageArea	1460	non-null	int64
62	GarageQual	1379	non-null	object
63	GarageCond	1379	non-null	object
64	PavedDrive	1460	non-null	object
65	WoodDeckSF	1460	non-null	int64
66	OpenPorchSF	1460	non-null	int64
67	EnclosedPorch	1460	non-null	int64
68	3SsnPorch	1460	non-null	int64
69	ScreenPorch	1460	non-null	int64
70	PoolArea	1460	non-null	int64
71	PoolQC	7	non-null	object
72	Fence	281	non-null	object

```
73 MiscFeature      54 non-null    object
74 MiscVal          1460 non-null   int64
75 MoSold           1460 non-null   int64
76 YrSold           1460 non-null   int64
77 SaleType         1460 non-null   object
78 SaleCondition     1460 non-null   object
79 SalePrice        1460 non-null   int64
dtypes: float64(3), int64(34), object(43)
memory usage: 912.6+ KB
```

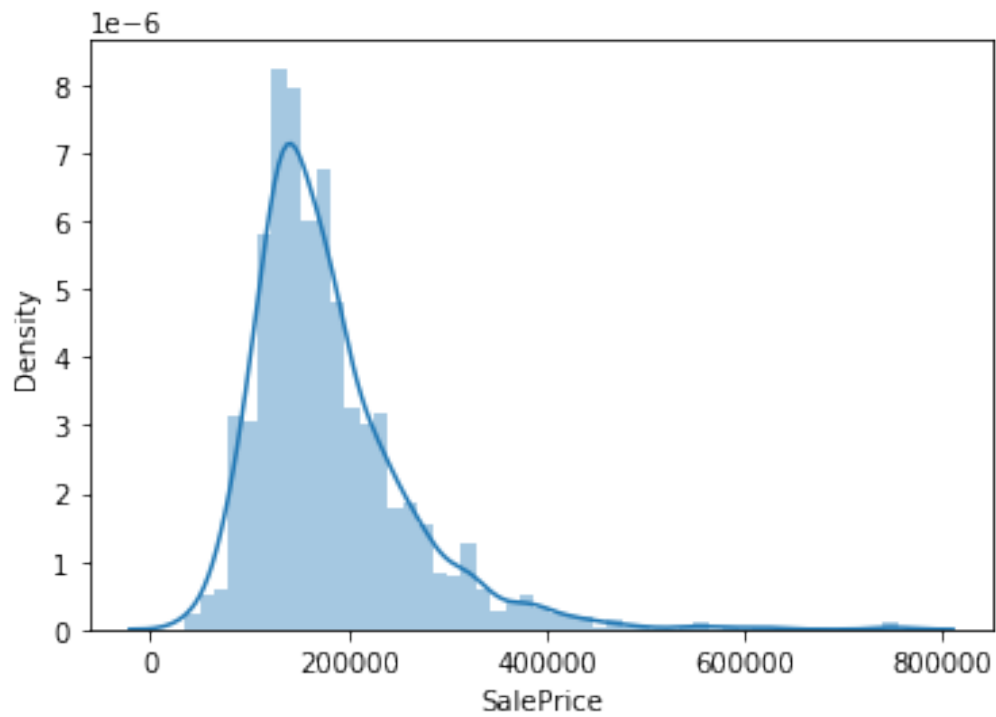
```
train["SalePrice"].describe()
```

```
count      1460.000000
mean       180921.195890
std        79442.502883
min        34900.000000
25%       129975.000000
50%       163000.000000
75%       214000.000000
max       755000.000000
Name: SalePrice, dtype: float64
```

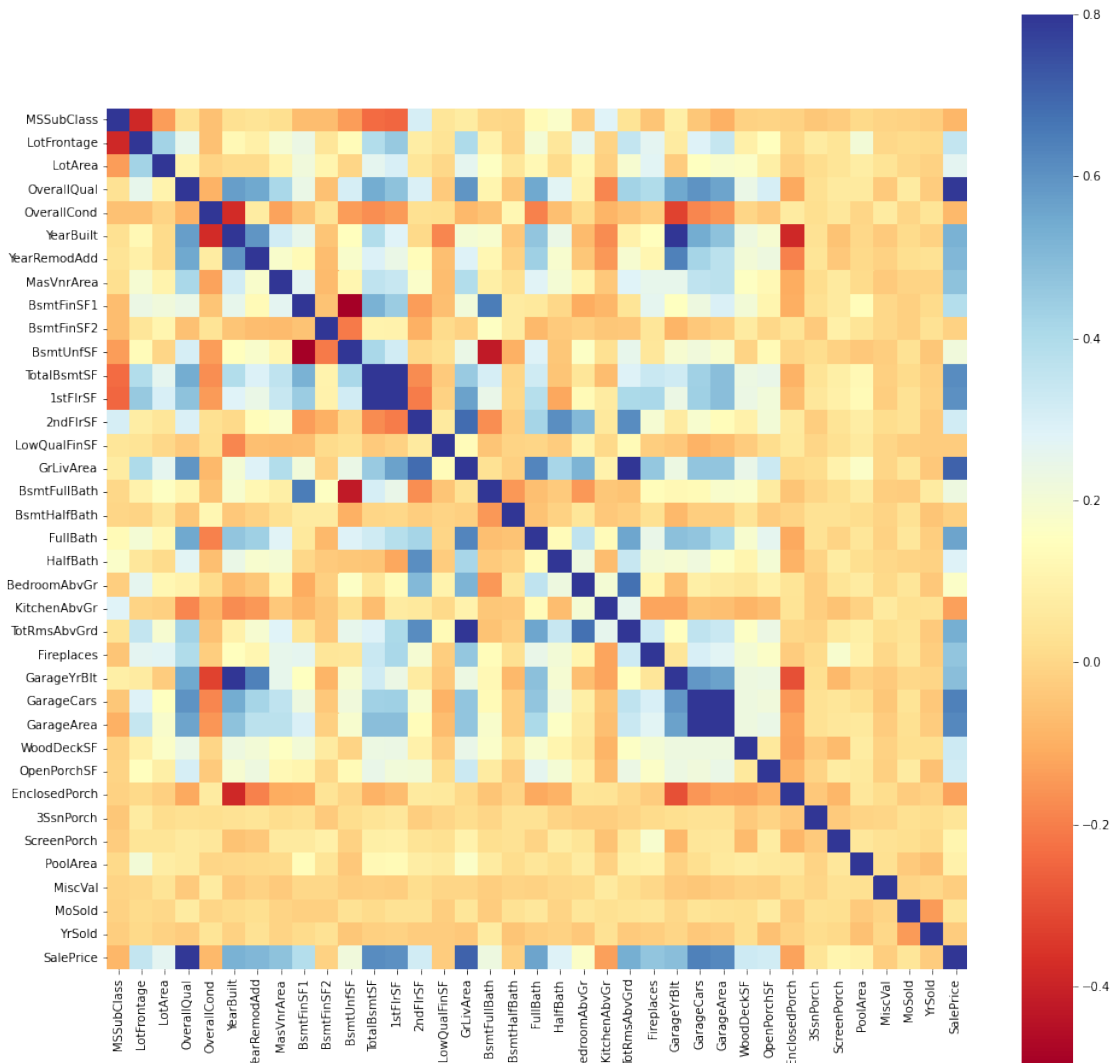
```
sns.distplot(train["SalePrice"])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619:
FutureWarning: `distplot` is a deprecated function and will be removed
in a future version. Please adapt your code to use either `displot` (a
figure-level function with similar flexibility) or `histplot` (an
axes-level function for histograms).
  warnings.warn(msg, FutureWarning)
```

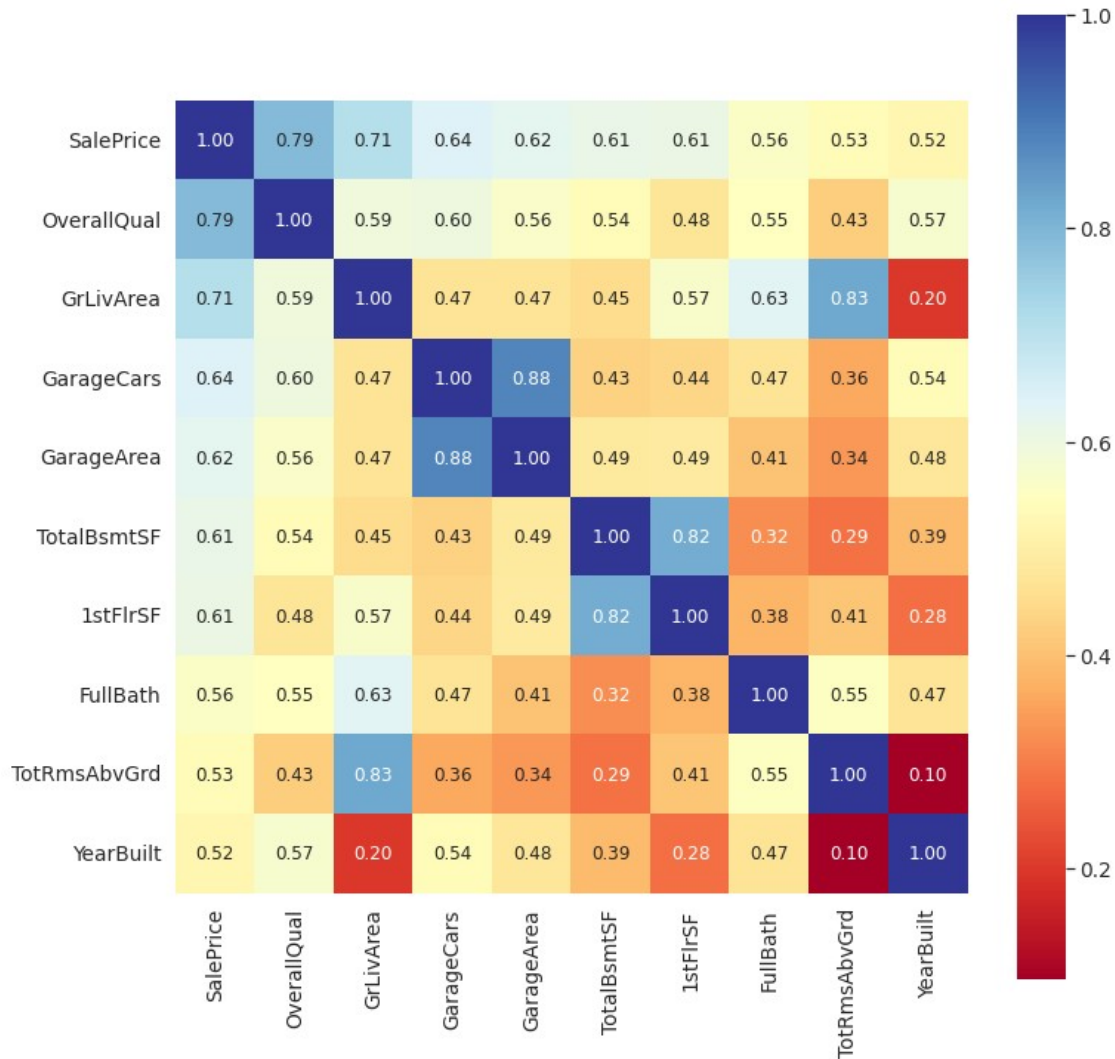
```
<matplotlib.axes._subplots.AxesSubplot at 0x7fefb09d5c10>
```



```
corrmat = train.corr(method='pearson')  
f, ax = plt.subplots(figsize=(16,16))  
sns.heatmap(corrmat,vmax=.8,square=True,cmap='RdYlBu')  
plt.show()
```

```
plt.figure(figsize=(12, 12))
k = 10 #Top k variabel yang berkorelasi dengan SalePrice
columns = corrmat.nlargest(k, 'SalePrice')['SalePrice'].index
cm = np.corrcoef(train[columns].values.T)
sns.set(font_scale=1.25)
hm = sns.heatmap(cm, cbar=True, annot=True, fmt='.2f', square=True,
annot_kws={'size': 12},
yticklabels=columns.values,
xticklabels=columns.values, cmap='RdYlBu')
plt.show()
```



```
train = train[train.GrLivArea < 4500]
```

```
total = test.isna().sum().sort_values(ascending=False)
```

```
#concatenate missing data into dataframe
```

```
missing = pd.concat([total],axis=1, keys=['Total'])
```

```
missing.head(45)
```

	Total
PoolQC	1456
MiscFeature	1408
Alley	1352
Fence	1169
FireplaceQu	730
LotFrontage	227
GarageYrBlt	78
GarageQual	78
GarageFinish	78
GarageCond	78
GarageType	76

BsmtCond	45
BsmtQual	44
BsmtExposure	44
BsmtFinType1	42
BsmtFinType2	42
MasVnrType	16
MasVnrArea	15
MSZoning	4
BsmtHalfBath	2
Utilities	2
Functional	2
BsmtFullBath	2
BsmtFinSF1	1
BsmtFinSF2	1
BsmtUnfSF	1
KitchenQual	1
TotalBsmtSF	1
Exterior2nd	1
GarageCars	1
Exterior1st	1
GarageArea	1
SaleType	1
MiscVal	0
BedroomAbvGr	0
KitchenAbvGr	0
YrSold	0
TotRmsAbvGrd	0
MoSold	0
Fireplaces	0
PoolArea	0
HalfBath	0
ScreenPorch	0
3SsnPorch	0
EnclosedPorch	0

```
train = train.drop(missing[missing.Total>0].index, axis=1)
```

```
test = test.dropna(axis=1)
```

```
test = test.drop(["Electrical"], axis=1)
```

Regression

```
predictor = ['OverallQual'] #X
```

```
out = ['SalePrice'] #(Yi)
```

```
#Rumus LR  $Y_i = b_0 + b_1X$ 
```

```
model = LinearRegression()
```

```
model.fit(train[predictor], train[out])
```

```
print(f'Intercept: {model.intercept_}') #Nilai untuk  $b_0$  atau c
```

```
print(f'Coefficient: {model.coef_[0]:}') #Nilai untuk  $b_1$  atau m
```

```

fitted = model.predict(train[predictor])  #(Yi hat=Ypred)
residuals = train[out] - fitted  #e=Yi-Yhat

Intercept: [-99155.50980987]
Coefficient: [45961.61275214]

ax = train.plot.scatter(x='OverallQual', y='SalePrice', figsize=(10,
7))
ax.plot(train.OverallQual, fitted, linewidth=5, color='k',
label=f'linear regression: Yi = {model.intercept_:} +
{model.coef_[0]}X')
for x, yactual, yfitted in zip(train.OverallQual, train.SalePrice,
fitted):
    ax.plot((x, x), (yactual, yfitted), '--', color='C1')
plt.tight_layout()
plt.legend()  #fungsi plt.legend () melacak gaya dan warna garis, dan
 mencocokkannya dengan label yang benar.
plt.show()

```

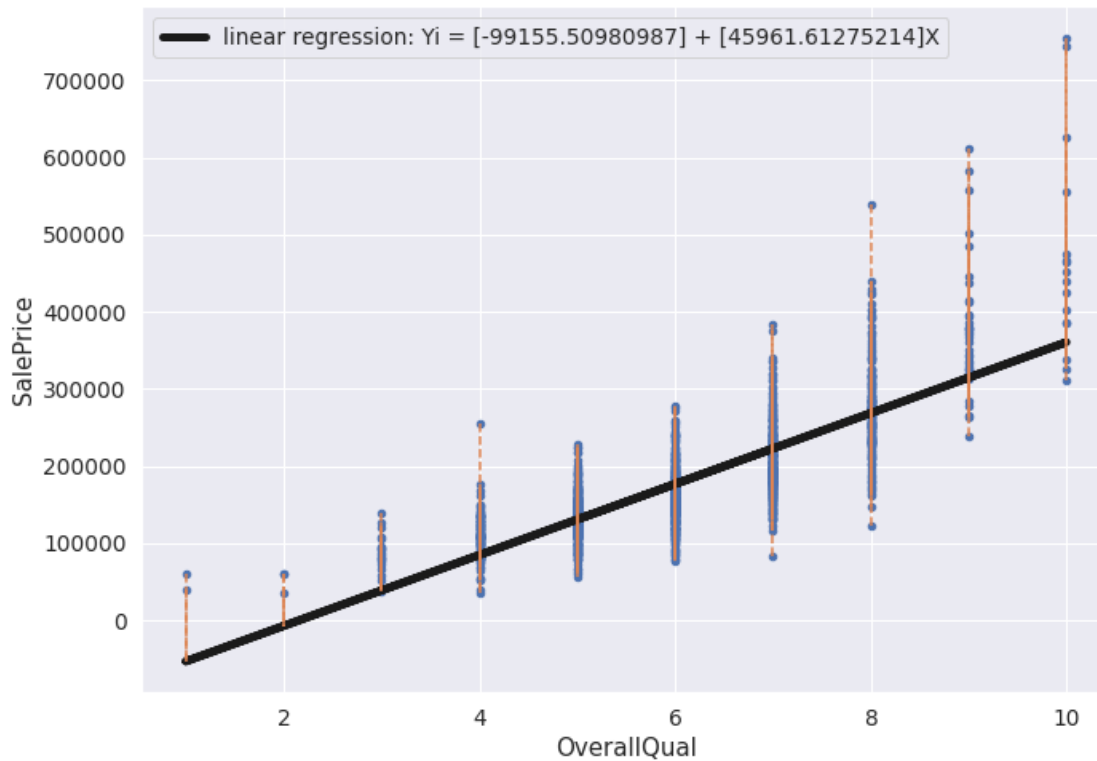
c argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with **x** & **y**. Please use the **color** keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.

/usr/local/lib/python3.7/dist-packages/numpy/core/shape_base.py:65: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

```

    ary = asanyarray(ary)

```



Prediction

```
full_df = pd.concat([train,test])
```

```
full_df = pd.get_dummies(full_df)
```

```
X = full_df.iloc[train.index]
```

```
X_test = full_df.iloc[test.index]
```

```
X = X.drop(['SalePrice'],axis=1)
```

```
X.shape
```

```
(1458, 154)
```

```
y = train.SalePrice
```

```
y.shape
```

```
(1458,)
```

```
from sklearn.model_selection import train_test_split
X_train, X_val, y_train, y_val = train_test_split(X, y,
train_size=0.8, random_state=42)
```

```
X.isna().sum().sort_values(ascending=False)
```

```
Id                                0
RoofMatl_CompShg                 0
HouseStyle_SLv1                 0
```

RoofStyle_Flat	0
RoofStyle_Gable	0
RoofStyle_Gambrel	0
RoofStyle_Hip	0
RoofStyle_Mansard	0
RoofStyle_Shed	0
RoofMatl_Membran	0
ExterQual_TA	0
RoofMatl_Metal	0
RoofMatl_Roll	0
RoofMatl_Tar&Grv	0
RoofMatl_WdShake	0
RoofMatl_WdShngl	0
ExterQual_Ex	0
ExterQual_Fa	0
HouseStyle_SFoyer	0
HouseStyle_2Story	0
HouseStyle_2.5Unf	0
HouseStyle_2.5Fin	0
Condition2_Feedr	0
Condition2_Norm	0
Condition2_PosA	0
Condition2_PosN	0
Condition2_RRAe	0
Condition2_RRAn	0
Condition2_RRNn	0
BldgType_1Fam	0
BldgType_2fmCon	0
BldgType_Duplex	0
BldgType_Twnhs	0
BldgType_TwnhsE	0
HouseStyle_1.5Fin	0
HouseStyle_1.5Unf	0
HouseStyle_1Story	0
ExterQual_Gd	0
ExterCond_Ex	0
MSSubClass	0
Electrical_SBrkr	0
HeatingQC_TA	0
CentralAir_N	0
CentralAir_Y	0
Electrical_FuseA	0
Electrical_FuseF	0
Electrical_FuseP	0
Electrical_Mix	0
PavedDrive_N	0
ExterCond_Fa	0
PavedDrive_P	0
PavedDrive_Y	0
SaleCondition_Abnorml	0

SaleCondition_AdjLand	0
SaleCondition_Alloca	0
SaleCondition_Family	0
SaleCondition_Normal	0
HeatingQC_Po	0
HeatingQC_Gd	0
HeatingQC_Fa	0
HeatingQC_Ex	0
ExterCond_Gd	0
ExterCond_Po	0
ExterCond_TA	0
Foundation_BrkTil	0
Foundation_CBlock	0
Foundation_PConc	0
Foundation_Slab	0
Foundation_Stone	0
Foundation_Wood	0
Heating_Floor	0
Heating_GasA	0
Heating_GasW	0
Heating_Grav	0
Heating_OthW	0
Heating_Wall	0
Condition2_Artery	0
Condition1_RRNn	0
Condition1_RRNe	0
LotShape_IR1	0
ScreenPorch	0
PoolArea	0
MiscVal	0
MoSold	0
YrSold	0
Street_Grvl	0
Street_Pave	0
LotShape_IR2	0
Condition1_RRAn	0
LotShape_IR3	0
LotShape_Reg	0
LandContour_Bnk	0
LandContour_HLS	0
LandContour_Low	0
LandContour_Lvl	0
LotConfig_Corner	0
3SsnPorch	0
EnclosedPorch	0
OpenPorchSF	0
WoodDeckSF	0
LotArea	0
OverallQual	0
OverallCond	0

YearBuilt	0
YearRemodAdd	0
1stFlrSF	0
2ndFlrSF	0
LowQualFinSF	0
GrLivArea	0
FullBath	0
HalfBath	0
BedroomAbvGr	0
KitchenAbvGr	0
TotRmsAbvGrd	0
Fireplaces	0
LotConfig_CulDSac	0
LotConfig_FR2	0
LotConfig_FR3	0
Neighborhood_NWAmes	0
Neighborhood_NridgHt	0
Neighborhood_OldTown	0
Neighborhood_SWISU	0
Neighborhood_Sawyer	0
Neighborhood_SawyerW	0
Neighborhood_Somerst	0
Neighborhood_StoneBr	0
Neighborhood_Timber	0
Neighborhood_Veenker	0
Condition1_Artery	0
Condition1_Feedr	0
Condition1_Norm	0
Condition1_PosA	0
Condition1_PosN	0
Condition1_RRAe	0
Neighborhood_NoRidge	0
Neighborhood_NPkVill	0
LotConfig_Inside	0
Neighborhood_NAmes	0
LandSlope_Gtl	0
LandSlope_Mod	0
LandSlope_Sev	0
Neighborhood_Blmngtn	0
Neighborhood_Blueste	0
Neighborhood_BrDale	0
Neighborhood_BrkSide	0
Neighborhood_ClearCr	0
Neighborhood_CollgCr	0
Neighborhood_Crawfor	0
Neighborhood_Edwards	0
Neighborhood_Gilbert	0
Neighborhood_IDOTRR	0
Neighborhood_MeadowV	0
Neighborhood_Mitchel	0


```
SaleCondition_Partial    0
dtype: int64

from sklearn.linear_model import LinearRegression
from scipy.stats import zscore
regressor = LinearRegression()
regressor.fit(X_train, y_train)
regressor.score(X_val, y_val)

0.014656716430643923

X_test = X_test.drop(["SalePrice"], axis=1)

y_preds = regressor.predict(X_test)
```