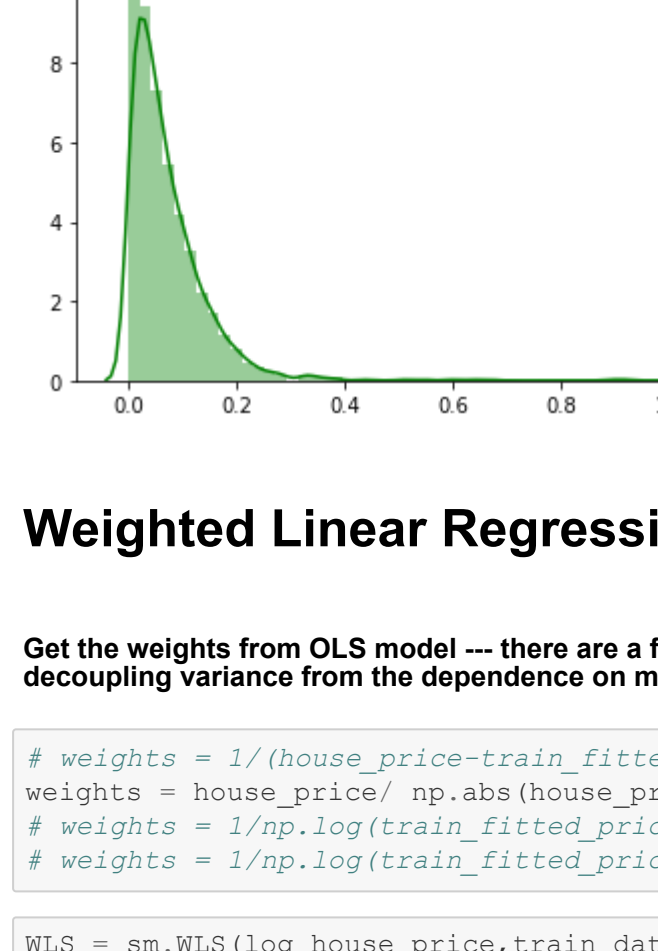



```
[38]: sns.distplot(weak_removed_train_percentage_error,color='g',label='train')
#sns.distplot(weak_removed_test_percentage_error,color='r',label='test')
plt.legend()
plt.title("Percentage Error after removing weak predictors")
```



Weighted Linear Regression Fit

Get the weights from OLS model --- there are a few common and possible ways to determine weights, all of which aims at decreasing variance from the dependence on mean

```
In [41]: # weights = 1/(house_price-train_fitted_price)**2
weights = house_price/ np.abs(house_price-train_fitted_price)
weights = 1/np.log(train_fitted_price)
# weights = 1/np.log(train_fitted_price)**2

In [43]: WLS = sm.WLS(log_house_price,train_data,weights=weights)

In [44]: weighted_lin_regression = WLS.fit()

In [45]: WLS_train_fitted_log_price = weighted_lin_regression.predict(train_data)
WLS_train_fitted_price = np.exp(WLS_train_fitted_log_price)
WLS_train_percentage_error = np.abs(house_price-WLS_train_fitted_price)/house_price
print('Average percentage error for training data: {}'.format(sum(WLS_train_percentage_error)/len(train_percentage_error)))

Average percentage error for training data: 0.0701636451716095

In [46]: print(weighted_lin_regression.summary())
```

YrSold	-0.0007	0.000	-1.616	0.106	-0.002	0.000
YrSold	0.0014	0.001	1.520	0.129	-0.000	0.003
LandContour_Lvl1	0.0004	0.006	0.724	0.469	-0.007	0.573
has_fence	-0.0014	0.003	-0.434	0.665	-0.008	0.005
gables_roof	-0.0065	0.003	-2.087	0.037	-0.013	-0.000
LandSlope_Gtl	0.0078	0.008	1.007	0.314	-0.007	0.023
Bath	0.0078	0.003	10.173	0.000	0.022	0.033
GarageScore	0.0082	0.001	5.607	0.000	0.005	0.011
OverallQual*OverallQual*GrLivArea	1.243e-06	1.78e-06	0.700	0.484	-2.24e-06	4.72e-06
OverallQual*GrLivArea	-1.039e-05	9.68e-06	-1.126	0.260	-2.39e-05	8.09e-06
OverallCond*GrLivArea	3.836e-06	1.11e-05	0.344	0.731	-1.8e-05	2.57e-05
OverallCond*OverallQual	-0.0024	0.003	-0.908	0.364	-0.008	0.003
ExteriorScore	-0.0040	0.003	-1.540	0.124	-0.009	0.003
BsmntLowQArea	0.0001	6.51e-06	15.369	0.000	8.73e-05	0.000
BsmntAvgQArea	0.0001	7.25e-06	18.696	0.000	0.000	0.000
BsmntHiQArea	0.0002	6.96e-06	26.219	0.000	0.000	0.000
RecentRemod	0.0042	0.004	1.185	0.236	-0.003	0.011
HouseAge	-0.0021	0.000	-14.771	0.000	-0.002	-0.002
MSZoning_Others	0.3518	0.090	3.910	0.000	0.175	0.528
MSZoning_RL	0.3973	0.090	4.438	0.000	0.222	0.573
MSZoning_RM	0.3782	0.087	4.254	0.000	0.203	0.553
lotShape_IR1	0.2816	0.067	4.178	0.000	0.149	0.414
lotShape_IR2	0.2855	0.068	4.229	0.000	0.153	0.418
lotShape_IR3	0.2816	0.068	4.140	0.000	0.148	0.415
lotShape_Reg	0.2786	0.067	4.136	0.000	0.146	0.411
lotConfig_Corner	0.3818	0.089	4.275	0.000	0.207	0.557
lotConfig_Inside	0.3688	0.090	4.116	0.000	0.193	0.545
lotConfig_Others	0.3766	0.090	4.202	0.000	0.201	0.552
Neighborhood_Blmngtn	0.0712	0.018	4.038	0.000	0.037	0.106
Neighborhood_Bluemtn	-0.0025	0.035	-0.072	0.943	-0.071	0.066
Neighborhood_Brdale	-0.0087	0.018	-0.487	0.626	-0.044	0.026
Neighborhood_BrkSlde	0.0616	0.015	4.172	0.000	0.033	0.091
Neighborhood_ClearCr	0.0623	0.016	3.832	0.000	0.030	0.094
Neighborhood_ColgrCr	0.0522	0.012	2.103	0.037	0.001	0.049
Neighborhood_Crawfor	0.1587	0.015	10.803	0.000	0.128	0.185
Neighborhood_Edwards	-0.0071	0.013	-0.547	0.585	-0.033	0.018
Neighborhood_Gilbert	0.2265	0.013	2.094	0.036	0.002	0.052
Neighborhood_Ingrht	-0.0232	0.016	-1.451	0.153	-0.055	0.005
Neighborhood_MeadowV	-0.1155	0.017	-6.620	0.000	-0.150	-0.081
Neighborhood_Mitchell	0.0096	0.012	0.781	0.435	-0.015	0.034
Neighborhood_Names	0.0357	0.012	2.907	0.002	0.013	0.058
Neighborhood_NPKVill	0.0826	0.021	3.951	0.000	0.042	0.124
Neighborhood_NWNames	0.0126	0.013	0.963	0.336	-0.013	0.038
Neighborhood_Noridge	0.0607	0.013	4.559	0.000	0.035	0.087
Neighborhood_NorHght	-0.1002	0.012	-8.771	0.000	-0.147	0.146
Neighborhood_OldTown	0.0092	0.014	0.651	0.515	-0.018	0.037
Neighborhood_SWISU	0.0582	0.018	3.182	0.001	0.022	0.094
Neighborhood_Sawyer	0.1916	0.058	3.319	0.000	0.073	0.311
Neighborhood_SawyerW	0.0338	0.012	2.768	0.006	0.010	0.058
Neighborhood_Somerst	0.1190	0.013	8.948	0.000	0.093	0.145
Neighborhood_Stonewr	0.1638	0.014	11.826	0.000	0.137	0.191
Neighborhood_Timber	0.0423	0.013	3.165	0.002	0.016	0.069
Neighborhood_Veenker	0.0841	0.015	5.557	0.000	0.054	0.114
Condition1_Artery	0.0845	0.031	2.685	0.007	0.023	0.146
Condition1_Feeder	0.1157	0.031	3.744	0.000	0.055	0.176
Condition1_Norm	0.1672	0.031	5.482	0.000	0.107	0.227
Condition1_PosA	0.1237	0.035	3.489	0.001	0.054	0.193
Condition1_PosB	0.1718	0.033	5.270	0.000	0.108	0.236
Condition1_RRAe	0.0543	0.034	1.592	0.112	-0.013	0.121
Condition1_RRAn	0.1466	0.031	4.682	0.000	0.085	0.208
Condition1_RRNe	0.1239	0.055	2.249	0.025	0.016	0.232
Condition1_RRNo	0.1385	0.034	4.044	0.000	0.064	0.214
Condition2_Artery	0.1665	0.051	3.271	0.001	0.067	0.266
Condition2_Feeder	0.1638	0.039	4.160	0.000	0.087	0.241
Condition2_Norm	0.1916	0.058	3.319	0.000	0.122	0.261
Condition2_PosA	0.3065	0.056	5.429	0.000	0.196	0.417
Condition2_PosB	0.0086	0.038	0.226	0.821	-0.066	0.083
Condition2_RRAe	0.0396	0.046	0.850	0.403	-0.040	0.100
Condition2_RRAn	0.1096	0.036	3.005	0.003	0.038	0.181
Condition2_RRNe	0.1711	0.061	2.814	0.005	0.052	0.290
BldgType_1Fam	0.2577	0.054	4.768	0.000	0.152	0.364
BldgType_2FmCon	0.1423	0.034	4.847	0.000	0.077	0.369
BldgType_Duplex	0.2297	0.055	4.159	0.000	0.122	0.338
BldgType_Twnhs	0.1720	0.055	3.116	0.002	0.064	0.280
BldgType_Twnhst	0.2048	0.054	3.756	0.000	0.099	0.311
HouseStyle_1.5Fin	0.1459	0.034	4.316	0.000	0.080	0.212
HouseStyle_1.5Unf	0.1703	0.035	4.875	0.000	0.102	0.239
HouseStyle_2FmCon	0.1536	0.034	4.500	0.000	0.087	0.221
HouseStyle_2.5Fin	0.1004	0.044	2.265	0.248	-0.035	0.136
HouseStyle_2.5Unf	0.1533	0.038	4.008	0.000	0.078	0.228
HouseStyle_2Story	0.1469	0.034	4.296	0.000	0.080	0.214
HouseStyle_3Story	0.1506	0.034	4.414	0.000	0.084	0.218
HouseStyle_Svl1	0.1562	0.034	4.574	0.000	0.089	0.223
MasVnrType_BrkFace	0.3713	0.089	4.153	0.000	0.196	0.547
MasVnrType_None	0.3771	0.090	4.209	0.000	0.201	0.553
MasVnrType_Others	0.3789	0.090	4.231	0.000	0.203	0.555
Foundation_BrkTl1	0.1931	0.045	4.262	0.000	0.104	0.282
Foundation_CBlock	0.2149	0.045	4.732	0.000	0.126	0.304
Foundation_PConc	0.2264	0.046	4.944	0.000	0.137	0.316
Foundation_Slab	0.2130	0.046	4.605	0.000	0.122	0.304
Foundation_Stone	0.1947	0.054	3.612	0.000	0.089	0.300
Foundation_Wood	0.0852	0.058	1.457	0.145	-0.029	0.200
Functional_Maj1	0.1959	0.041	4.792	0.000	0.116	0.276
Functional_Maj2	0.0569	0.049	1.162	0.246	-0.039	0.153
Functional_Min1	0.2785	0.040	6.970	0.000	0.200	0.357
Functional_Min2	0.2620	0.040	6.533	0.000	0.183	0.341
Functional_Svl1	0.1716	0.040	4.276	0.000	0.093	0.250
Functional_Sev	-0.1396	0.041	-3.416	0.001	-0.220	-0.059
Functional_Type	0.3019	0.038	7.847	0.000	0.226	0.377
GarageType_Attchd	0.3771	0.090	4.210	0.000	0.201	0.553
GarageType_Detachd	0.3779	0.089	4.240	0.000	0.203	0.553
GarageType_Others	0.3722	0.090	4.142	0.000	0.196	0.548
SaleType_COD	0.0598	0.034	1.778	0.076	-0.006	0.126
SaleType_CND	0.1718	0.058	2.951	0.003	0.058	0.286
SaleType_Con	0.2078	0.046	4.491	0.000	0.117	0.299
SaleType_ConLd	0.1472	0.036	4.105	0.000	0.077	0.218
SaleType_ConLr	0.0495	0.042	1.190	0.234	-0.032	0.131
SaleType_ConLs	0.0591	0.035	1.681	0.093	-0.010	0.128
SaleType_New	0.1784	0.067	2.690	0.007	0.049	0.310
SaleType_Oth	0.1772	0.054	3.264	0.001	0.071	0.284
SaleType_Web	0.0755	0.032	2.328	0.020	0.012	0.139
SaleCondition_Abnormal	-0.0077	0.068	-0.114	0.909	-0.141	0.126
SaleCondition_AdjLend	0.0559	0.073	0.775	0.439	-0.086	0.199
SaleCondition_AdLoca	0.0634	0.073	0.873	0.383	-0.080	0.207
SaleCondition_Family	-0.0008	0.070	0.013	0.980	-0.136	0.137
SaleCondition_Normal	0.0559	0.068	0.824	0.410	-0.077	0.189
Exterior_HdBoard	0.1744	0.045	3.862	0.000	0.085	0.260
Exterior_MetalSd	0.1997	0.045	4.465	0.000	0.112	0.287
Exterior_Others	0.2063	0.045	4.577	0.000	0.118	0.295
Exterior_Plywood	0.1789	0.045	3.946	0.000	0.090	0.268
Exterior_VinylSd3	0.1950	0.045	4.338	0.000	0.107	0.283
Exterior_Wd Sng	0.1750	0.045	3.921	0.000	0.087	0.263
Omnibus:	249.570	Durbin-Watson:	1.895			
Prob (Omnibus):	0.000	Jarque-Bera (JB):	55.479			
Skew:	-0.052	Prob (JB):	8.97e-13			
Kurtosis:	2.047	Cond. No.	1.16e+16			
=====						
Warnings:						
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.						
[2] The condition number is large, 1.16e+16. This might indicate that there are strong multicollinearity or other numerical problems.						
=====						
WLS_prediction = pd.DataFrame(np.exp(weighted_lin_regression.predict(test_data)))						
WLS_prediction = WLS_prediction.rename(columns={'SalePrice'})						
=====						
WLS_prediction.to_csv('E:/projects/housing_price/WLS_prediction.csv', columns=WLS_prediction.columns)						
=====						