Bandwidth_visualization

May 20, 2020

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In [ ]: #@title Python imports
        # A bit of imports
        import numpy as np
        import pandas as pd
        import warnings
        warnings.filterwarnings('ignore')
        import numpy as np
        import pandas as pd
        import ipywidgets as widgets
        from ipywidgets import interact, interact_manual
        from matplotlib import pyplot as plt
        import seaborn as sns
        import warnings
        warnings.filterwarnings("ignore")
        from scipy import stats
        import geopandas as gpd
        from shapely.geometry import Point, Polygon
        %matplotlib inline
        sns.set(color_codes=True)
        from sklearn import linear model
        from matplotlib.colors import LinearSegmentedColormap
        from matplotlib_scalebar.scalebar import ScaleBar
        import statsmodels.api as statm
        import libpysal as ps
        from mgwr.gwr import GWR
        from mgwr.gwr import MGWR
        from mgwr.sel_bw import Sel_BW
        from spglm.family import Gaussian, Binomial, Poisson
        import multiprocessing as mp
        pool = mp.Pool()
        import io
In [23]: df = pd.read_csv("example_dataset.csv")
In [24]: df.columns
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Out [24]: Index(['the_geom', 'longitude', 'the_geom_webmercator', 'avg_tech', 'latitude',
                                       'houses', 'avg_unemp', 'cartodb_id', 'avg_index', 'avg_price',
                                       'avg_basement', 'avg_sqft', 'avg_water_dist', 'avg_age', 'ind'],
                                    dtype='object')
In [26]: df['ln_price']=np.log(df['avg_price'])
In [29]: df['ln_price']
Out[29]: 0
                                       12.611636
                      1
                                      12.312879
                      2
                                       12.563657
                      3
                                      12.715962
                                      12.619820
                                      12.936781
                     354
                                      12.733797
                     355
                      356
                                      12.839368
                      357
                                      13.007454
                     358
                                      13.426880
                     Name: ln_price, Length: 359, dtype: float64
In [40]: df=df.dropna()
In [42]: import statsmodels.api as statm
                     X=df[['avg_tech','avg_unemp','avg_index','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_basement','avg_water_dist','avg_sqft','avg_sqft','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','avg_water_dist','a
                     X_std = (X-X.mean(axis=0))/X.std(axis=0)
                     X_std=statm.add_constant(X_std)
                     y=df['ln_price']
                     y_std = (y-y.mean(axis=0))/y.std(axis=0)
                     model = statm.OLS(y_std, X_std).fit()
                     predictions=model.predict(X_std)
                     model.summary()
Out[42]: <class 'statsmodels.iolib.summary.Summary'>
                                                                                          OLS Regression Results
                      ______
                     Dep. Variable:
                                                                                             ln_price
                                                                                                                       R-squared:
                                                                                                                                                                                                        0.898
                     Model:
                                                                                                        OLS Adj. R-squared:
                                                                                                                                                                                                        0.896
                     Method:
                                                                                Least Squares F-statistic:
                                                                                                                                                                                                        438.4
                     Date:
                                                                        Wed, 20 May 2020 Prob (F-statistic):
                                                                                                                                                                                           1.15e-168
                     Time:
                                                                                             12:47:31 Log-Likelihood:
                                                                                                                                                                                                   -98.782
                     No. Observations:
                                                                                                         357 AIC:
                                                                                                                                                                                                        213.6
                     Df Residuals:
                                                                                                         349
                                                                                                                       BIC:
                                                                                                                                                                                                        244.6
                     Df Model:
                                                                                                             7
                     Covariance Type:
                                                                                         nonrobust
```

		coef	std err	t	P> t	[0.025	0.975]
	const	6.721e-15	0.017	3.93e-13	1.000	-0.034	0.034
	avg_tech	0.2186	0.029	7.433	0.000	0.161	0.276
	avg_unemp	-0.2990	0.028	-10.802	0.000	-0.353	-0.245
	avg_index	-0.1231	0.030	-4.115	0.000	-0.182	-0.064
	avg_sqft	0.5865	0.023	25.043	0.000	0.540	0.633
	avg_basement	0.0984	0.023	4.212	0.000	0.052	0.144
	avg_water_dist	-0.2106	0.034	-6.280	0.000	-0.277	-0.145
	avg_age	0.1563	0.025	6.371	0.000	0.108	0.205
	Omnibus:17.936Durbin-Watson:Prob(Omnibus):0.000Jarque-Bera (JB):		son:	1.793			
			0.000) Jarque-Bera (JB):		44.569	
	Skew:		0.139	Prob(JB):		2.10e-10	
	Kurtosis:		4.708 	Cond. No.	========	.=======	4.47 =====
In [43]:	<pre>Warnings: [1] Standard Errors assume that the covariance matrix of the errors is correctly spec """ : import libpysal as ps from mgwr.gwr import GWR from mgwr.gwr import MGWR from mgwr.sel_bw import Sel_BW from spglm.family import Gaussian, Binomial, Poisson import multiprocessing as mp pool = mp.Pool()</pre>						
In [44]:	df.columns						
Out[44]:	<pre>: Index(['the_geom', 'longitude', 'the_geom_webmercator', 'avg_tech', 'latitude',</pre>						
In [45]:	<pre>coords = np.array(list(zip(df['longitude'],df['latitude']))) y = np.array(df['ln_price']).reshape((-1,1)) y_std = (y-y.mean(axis=0))/y.std(axis=0) X=df[['avg_tech','avg_unemp','avg_index','avg_sqft','avg_basement','avg_water_dist','xy_std=(X-X.mean(axis=0))/X.std(axis=0) selector_gwr = Sel_BW(coords, y_std, X_std)</pre>						
In [48]:	<pre>selector_mgwr = Sel_BW(coords, y_std, X_std, multi=True)</pre>						
In [49]:	selector_mgwr.search(pool=pool)						
Out[49]:	array([43., 16	4., 226., 19	1., 45.,	81., 45., 3	54.])		