Tests with Geopandas

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In [2]:

import geopandas as gpd
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline

In [6]:

nybb_path = gpd.datasets.get_path('nybb')
nybb_path

Out[6]:

 $\label{lib} $$ 'zip://C:\ProgramData\Anaconda3\\lib\\site-packages\\geopandas\\datasets\\nybb_16a.zip'$

In [8]:

boros = gpd.read_file(nybb_path)
boros.set_index('BoroCode', inplace=True)

Out[8]:

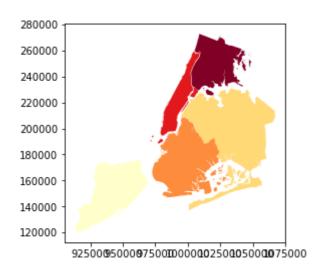
	BoroName	Shape_Leng	Shape_Area	geometry
BoroCode				
5	Staten Island	330470.010332	1.623820e+09	(POLYGON ((970217.0223999023 145643.3322143555
4	Queens	896344.047763	3.045213e+09	(POLYGON ((1029606.076599121 156073.8142089844
3	Brooklyn	741080.523166	1.937479e+09	(POLYGON ((1021176.479003906 151374.7969970703
1	Manhattan	359299.096471	6.364715e+08	(POLYGON ((981219.0557861328 188655.3157958984
2	Bronx	464392.991824	1.186925e+09	(POLYGON ((1012821.805786133 229228.2645874023

In [15]:

boros.plot(cmap='YlOrRd')

Out[15]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac75be3c8>



In [18]:

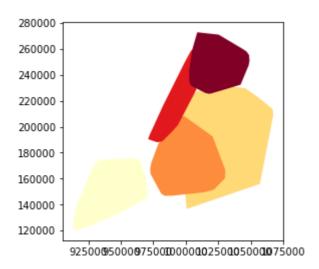
boros_x=boros['geometry'].convex_hull

In [19]:

boros_x.plot(cmap='YlOrRd')

Out[19]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac83f4908>



With Portugal Map

In [9]:

path = 'ML/GeoPandas/Portugal/PRT_adm_shp/'

In [12]:

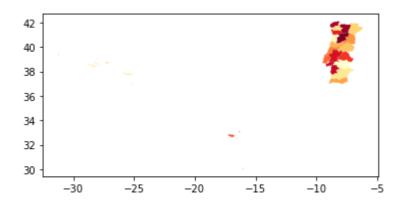
```
# Tables Portugal / Distritos / Concelhos
uf_pt1 = gpd.read_file(path + 'PRT_adm0.shp')
uf_pt2 = gpd.read_file(path + 'PRT_adm1.shp')
uf_pt3 = gpd.read_file(path + 'PRT_adm2.shp')
```

In [13]:

```
# Distritos
uf_pt2.plot(cmap='YlOrRd')
```

Out[13]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac74274e0>



In [16]:

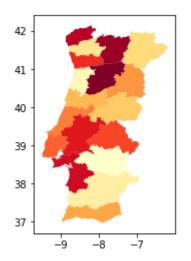
```
# Delete information about Autonomos Regions (Açores e Madeira)
uf_pt2x = uf_pt2[uf_pt2.TYPE_1 != 'Regiões autônoma']
```

In [17]:

```
uf_pt2x.plot(cmap='YlOrRd')
```

Out[17]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac74285c0>



In [20]:

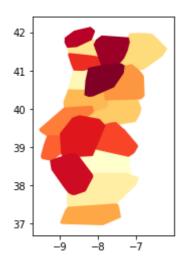
```
uf_pt2x_x=uf_pt2x['geometry'].convex_hull
```

In [21]:

```
uf_pt2x_x.plot(cmap='Y10rRd')
```

Out[21]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac7e4f278>



In [22]:

```
world = gpd.read_file(gpd.datasets.get_path('naturalearth_lowres'))
cities = gpd.read_file(gpd.datasets.get_path('naturalearth_cities'))
```

In [27]:

world.head()

Out[27]:

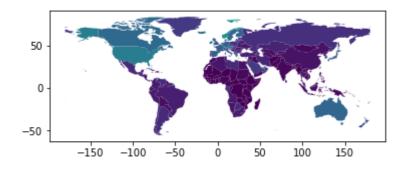
							_
	pop_est	continent	name	iso_a3	gdp_md_est	geometry	ć
0	28400000.0	Asia	Afghanistan	AFG	22270.0	POLYGON ((61.21081709172574 35.65007233330923,	С
1	12799293.0	Africa	Angola	AGO	110300.0	(POLYGON ((16.32652835456705 -5.87747039146621	С
2	3639453.0	Europe	Albania	ALB	21810.0	POLYGON ((20.59024743010491 41.85540416113361,	С
3	4798491.0	Asia	United Arab Emirates	ARE	184300.0	POLYGON ((51.57951867046327 24.24549713795111,	С
4	40913584.0	South America	Argentina	ARG	573900.0	(POLYGON ((-65.500000000000003 -55.1999999999999	C

In [26]:

Simply use the plot command with the column argument set to the column whose values y
ou want used to assign colors.
#world = world[(world.pop_est>0) & (world.name!="Antarctica")]
world['gdp_per_cap'] = world.gdp_md_est / world.pop_est
world.plot(column='gdp_per_cap')

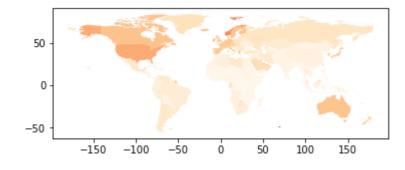
Out[26]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ac7893cf8>



In [28]:

world.plot(column='gdp_per_cap', cmap='OrRd');

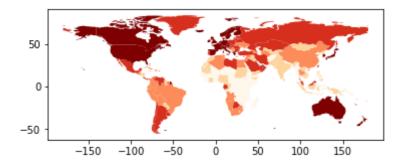


In [30]:

The scheme option can be set to 'equal_interval', 'quantiles' or 'percentiles'.
world.plot(column='gdp_per_cap', cmap='OrRd', scheme='quantiles')

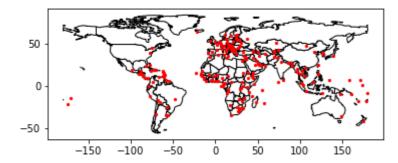
Out[30]:

<matplotlib.axes._subplots.AxesSubplot at 0x16acaa7a668>



In [32]:

```
# set aspect to equal. This is done automatically
# when using *geopandas* plot on it's own, but not when
# working with pyplot directly.
fig, ax = plt.subplots()
ax.set_aspect('equal')
world.plot(ax=ax, color='white', edgecolor='black')
cities.plot(ax=ax, marker='o', color='red', markersize=5)
plt.show();
```



London Maps

In [35]:

```
# Map of crime in 2016 - London
# 2016-01-metropolitan-street
# https://data.london.gov.uk/dataset/london-borough-profiles
path_london = 'ML/GeoPandas/London/'
```

In [38]:

```
# set the filepath and load in a shapefile
fp = path_london + 'London_Borough_Excluding_MHW.shp'
map_df = gpd.read_file(fp)
# check data type so we can see that this is not a normal dataframe, but a GEOdataframe
map_df.head()
```

Out[38]:

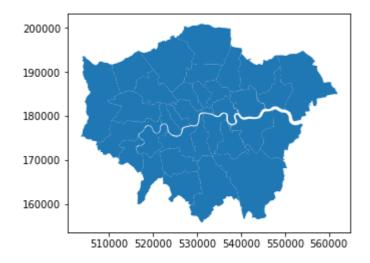
	NAME	GSS_CODE	HECTARES	NONLD_AREA	ONS_INNER	SUB_2009	SUB_2
0	Kingston upon Thames	E09000021	3726.117	0.000	F	None	None
1	Croydon	E09000008	8649.441	0.000	F	None	None
2	Bromley	E09000006	15013.487	0.000	F	None	None
3	Hounslow	E09000018	5658.541	60.755	F	None	None
4	Ealing	E09000009	5554.428	0.000	F	None	None

In [39]:

map_df.plot()

Out[39]:

<matplotlib.axes._subplots.AxesSubplot at 0x16acab0b860>



In [51]:

```
# Read Excel file
df = pd.read_excel(path_london + 'london-borough-profiles.xls')
df.head(10)
```

Out[51]:

	Code	Area_name	Inner/_Outer_London	GLA_Population_Estimate_2017	GL
0	E0900001	City of London	Inner London	8800	532
1	E09000002	Barking and Dagenham	Outer London	209000	78 ⁻
2	E0900003	Barnet	Outer London	389600	15 ⁻
3	E09000004	Bexley	Outer London	244300	97
4	E09000005	Brent	Outer London	332100	12 ⁻
5	E09000006	Bromley	Outer London	327900	14(
6	E09000007	Camden	Inner London	242500	10
7	E09000008	Croydon	Outer London	386500	159
8	E09000009	Ealing	Outer London	351600	13:
9	E09000010	Enfield	Outer London	333000	130

10 rows × 84 columns

→

In [44]:

```
# Create Maps online
```

https://app.datawrapper.de/map/EuqvD/basemap

 ${\it \# https://towards} datascience.com/lets-make-a-map-using-geopandas-pandas-and-matplotlib-to-make-a-chloropleth-map-dddc31c1983d$

In [60]:

In [61]:

```
df1.head()
```

Out[61]:

	Area_name	GLA_Population_Estimate_2017	Population_density_(per_hectare)_20
0	City of London	8800	30.3
1	Barking and Dagenham	209000	57.9
2	Barnet	389600	44.9
3	Bexley	244300	40.3
4	Brent	332100	76.8
4			+

In [65]:

Out[65]:

	Area_name	GLA_Population	Density_per_hectare	Mortality_rate
0	City of London	8800	30.3	129
1	Barking and Dagenham	209000	57.9	228
2	Barnet	389600	44.9	134
3	Bexley	244300	40.3	164
4	Brent	332100	76.8	169

In [67]:

```
# join the geodataframe with the cleaned up csv dataframe
merged = map_df.set_index('NAME').join(data_for_map.set_index('Area_name'))
merged.head()
```

Out[67]:

	GSS_CODE	HECTARES	NONLD_AREA	ONS_INNER	SUB_2009	SUB_20
NAME						
Kingston upon Thames	E09000021	3726.117	0.000	F	None	None
Croydon	E09000008	8649.441	0.000	F	None	None
Bromley	E09000006	15013.487	0.000	F	None	None
Hounslow	E09000018	5658.541	60.755	F	None	None
Ealing	E09000009	5554.428	0.000	F	None	None

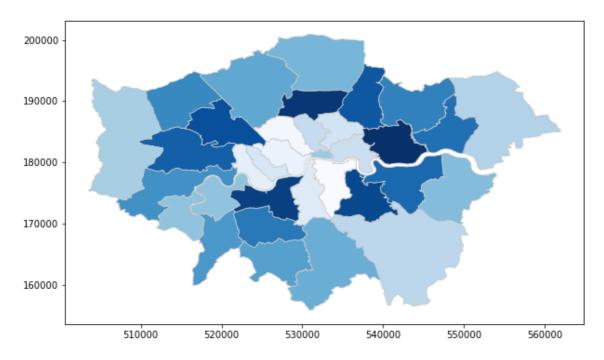
Map time!

In [69]:

```
# set a variable that will call whatever column we want to visualise on the map
variable = 'Density_per_hectare'
# set the range for the choropleth
vmin, vmax = 120, 220
# create figure and axes for Matplotlib
fig, ax = plt.subplots(1, figsize=(10, 6))
# create map
merged.plot(column=variable, cmap='Blues', linewidth=0.8, ax=ax, edgecolor='0.8')
```

Out[69]:

<matplotlib.axes._subplots.AxesSubplot at 0x16ad2097198>



Customising the map

In [73]:

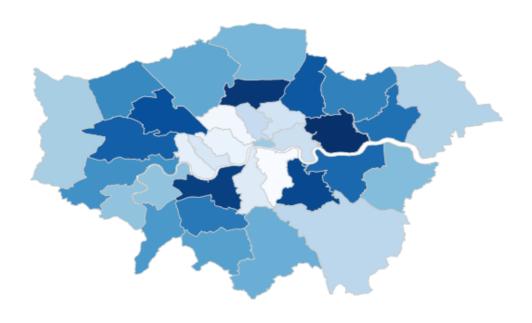
REMOVE THE AXIS
ax.axis('off')

In [72]:

```
# set a variable that will call whatever column we want to visualise on the map
variable = 'Density_per_hectare'
# set the range for the choropleth
vmin, vmax = 120, 220
# create figure and axes for Matplotlib
fig, ax = plt.subplots(1, figsize=(10, 6))
# remove the axis
ax.axis('off')
# create map
merged.plot(column=variable, cmap='Blues', linewidth=0.8, ax=ax, edgecolor='0.8')
```

Out[72]:

<matplotlib.axes._subplots.AxesSubplot at 0x16aca7058d0>



In [80]:

ADD A TITLE

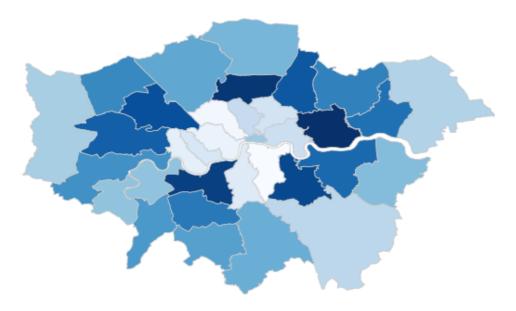
In [79]:

```
# set a variable that will call whatever column we want to visualise on the map
variable = 'Density_per_hectare'
# set the range for the choropleth
vmin, vmax = 120, 220
# create figure and axes for Matplotlib
fig, ax = plt.subplots(1, figsize=(10, 6))
# remove the axis
ax.axis('off')
# create map
merged.plot(column=variable, cmap='Blues', linewidth=0.8, ax=ax, edgecolor='0.8')
# add a title
ax.set_title('Preventable death rate in London', fontdict={'fontsize': '25', 'fontweigh'
t': '3'})
# create an annotation for the data source
ax.annotate('Source: London Datastore, 2014',xy=(0.1, .08), xycoords='figure fraction'
            horizontalalignment='left', verticalalignment='top', fontsize=12, color='#5
55555')
```

Out[79]:

<matplotlib.text.Annotation at 0x16acaacac18>

Preventable death rate in London



Source: London Datastore, 2014

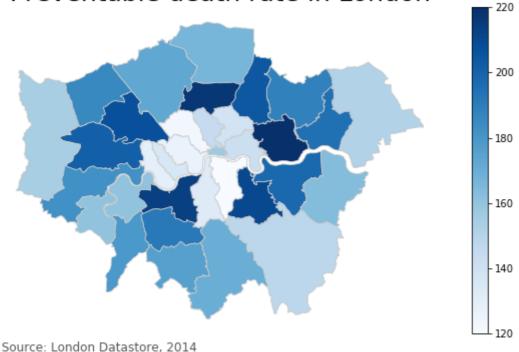
In [76]:

Create colorbar as a legend

In [78]:

```
# set a variable that will call whatever column we want to visualise on the map
variable = 'Density_per_hectare'
# set the range for the choropleth
vmin, vmax = 120, 220
# create figure and axes for Matplotlib
fig, ax = plt.subplots(1, figsize=(10, 6))
# remove the axis
ax.axis('off')
# create map
merged.plot(column=variable, cmap='Blues', linewidth=0.8, ax=ax, edgecolor='0.8')
# add a title
ax.set_title('Preventable death rate in London', fontdict={'fontsize': '25', 'fontweigh
t': '3'})
# create an annotation for the data source
ax.annotate('Source: London Datastore, 2014',xy=(0.1, .08), xycoords='figure fraction'
            horizontalalignment='left', verticalalignment='top', fontsize=12, color='#5
55555')
# Create colorbar as a legend
sm = plt.cm.ScalarMappable(cmap='Blues', norm=plt.Normalize(vmin=vmin, vmax=vmax))
# empty array for the data range
sm._A = []
# add the colorbar to the figure
cbar = fig.colorbar(sm)
```





In [82]:

```
# Save figure
# The code below will save the figure as a png
```

In	[84]

fig.savefig('map_export.png', dpi=300)

In []: