I. Library import

```
In [1]: #Import the librairies
        import ipywidgets as widgets
        from ipywidgets import interact, interactive, HBox, VBox
        import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        import geopandas as gpd
        import geodatasets
        import matplotlib.pyplot as plt
        import geopandas as gpd
        import pandas as pd
        import folium
        import matplotlib
        import mapclassify
        from folium.plugins import MarkerCluster
        import os
In [2]: #Read the data from 2024 and 2023 and concat them
        data2 = pd.read_csv("data.csv",low_memory=False)
        data1 = pd.read_csv("data2023.csv",low_memory=False)
        data = pd.concat([data1, data2], ignore_index=True)
In [3]: data.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 1584106 entries, 0 to 1584105
      Data columns (total 18 columns):
       # Column
                                     Non-Null Count
                                                       Dtype
       ---
                                                       ----
                                     -----
       0 date_mutation
                                    1584106 non-null object
                                    1584106 non-null object
       1
           nature mutation
                                   1584106 non-null float64
       2 valeur_fonciere
       3 code commune
                                    1584106 non-null object
                                   1584106 non-null object
       4 code_departement
                                    1584106 non-null object
       5
           id parcelle
       6 nombre_lots
                                    1584106 non-null int64
       7
                                     1584106 non-null float64
           code type local
           nombre_pieces_principales 1584106 non-null float64
                                     1584106 non-null float64
       9
           surface terrain
       10 longitude
                                    1584106 non-null float64
       11 latitude
                                    1584106 non-null float64
       12 tranche_valeur
                                    1584106 non-null object
                                    1584106 non-null int64
       13 code insee
       14 region name
                                    1584106 non-null object
                                    1584106 non-null object
       15 plage_surface
                                     1584106 non-null float64
       16 prix m2
       17 mois_annee
                                     1584106 non-null object
       dtypes: float64(7), int64(2), object(9)
      memory usage: 217.5+ MB
```

II. Création des fonctions

```
In [4]: def selection_filters_and_analyses(df):
    # Presentation of the filters
```

```
type_local_map = {
    1: 'House',
    2: 'Apartment',
    3: 'Dependency (Isolated)',
    4: 'Industrial and commercial premises or similar'
df['type_local'] = df['code_type_local'].map(type_local_map)
departments = df['code_departement'].unique()
types_properties = df['type_local'].unique()
# Widgets for the filters
department_select = widgets.SelectMultiple(
    options=departments,
    value=[],
    description='Departments',
    layout=widgets.Layout(width='60%')
department select.style.description width = '100px'
types_properties = df['type_local'].unique()
types_properties_select = widgets.SelectMultiple(
    options=types_properties,
    value=[],
    description='Types of property',
    layout=widgets.Layout(width='60%')
types_properties_select.style.description_width = '100px'
price range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150k - 200k €",
price_range_select = widgets.SelectMultiple(
    options=price_range,
    value=[],
    description='Price range',
    layout=widgets.Layout(width='60%')
price_range_select.style.description_width = '100px'
slice_surface = ['0-50 \text{ m}^2', '50-100 \text{ m}^2', '100-150 \text{ m}^2', '150-200 \text{ m}^2', '200-25]
slice surface select = widgets.SelectMultiple(
    options=slice_surface,
    value=[],
    description='Surface slice',
    layout=widgets.Layout(width='60%')
slice surface select.style.description width = '100px'
# Widget to sekect the analyse to execute
disponible_analyses = {
    "Average price per m2": show_average_price_per_m2,
    "Average surface area" : show_average_surface_per_dept,
    "Price evolution" : evolution average price per m2 per departement month
    "Number of main parts" :display_number_of_room_department,
    "Type of property by department" : distribution_type_property_by_departm
    "Distribution of land values" : distribution_of_land_value_by_tranche,
    "Price per m<sup>2</sup> depending on the surface area" : price_m2_per_surface,
    "Explore the map" : explore_france
}
```

analyse_select = widgets.SelectMultiple(

```
options=list(disponible_analyses.keys()),
                description='Analyse',
                layout=widgets.Layout(width='60%')
            analyse_select.style.description_width = '100px'
            # Button to apply the filters
            bouton_apply = widgets.Button(description="Apply the filters", button_style=
            # Function to take the filters and execute the analyse
            def apply_filter(_):
                selected_analyses = list(analyse_select.value)
                # We execute each analyse selected
                for analyse in selected_analyses:
                    func = disponible analyses[analyse]
                    func(df, department_select, types_properties_select,price_range_sele
            # Link the button to the function
            bouton_apply.on_click(apply_filter)
            # Show widgets
            display(department_select, types_properties_select,price_range_select,slice_
In [5]: def show_average_price_per_m2(df,department_select,types_properties_select,price
            # We take the department, price range, slice surface and types of properties
            selected_departments = list(department_select.value)
            selected_types_properties = list(types_properties_select.value)
            selected_price_range = list(price_range_select.value)
            selected_slice_surface = list(slice_surface_select.value)
            # If nothing is selected we take all the values for each point
            if not selected departments:
                selected_departments = list(df['code_departement'].unique())
            if not selected types properties:
                selected types properties = list(df['type local'].unique())
            if not selected_price_range:
                selected price range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150
            if not selected_slice_surface:
                selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
            # Filter data based with our filter
            df_filtered = df[(df['code_departement'].isin(selected_departments)) &
                                 (df['type_local'].isin(selected_types_properties))&
                                 (df['tranche_valeur'].isin(selected_price_range))&
                                 (df['plage_surface'].isin(selected_slice_surface))].copy
            df_filtered_national = df[
                                 (df['type local'].isin(selected types properties))&
                                 (df['tranche_valeur'].isin(selected_price_range))&
                                 (df['plage_surface'].isin(selected_slice_surface))].copy
            # Calculate the average price per m<sup>2</sup> by department
            average_price_par_dept = df_filtered.groupby('code_departement')['prix_m2'].
            # Calculate the average price of the selected departments
```

```
average_price_select = average_price_par_dept.mean()
# Calculate the national average price per m<sup>2</sup>
national_average_price = df_filtered_national[df_filtered_national['type_loc
# we create the graph
fig, ax = plt.subplots(figsize=(20, 6))
# We draw a bar to show the average price for each department
ax.bar(average_price_par_dept.index, average_price_par_dept)
# Added a bar for the average price selected
ax.bar('Selected Average', average_price_select, alpha=0.7)
# Added a bar for the national average price
ax.bar('National Average', national_average_price, alpha=0.7)
# And this is the parameters of the graphics
ax.set_ylabel('Average price per m² (€)')
ax.set_xlabel('Departments')
ax.set_title("Average price per m² per department, selected and glogal avera
ax.grid(True, linestyle='--', alpha=0.6)
plt.xticks(rotation=90)
plt.tight_layout() # To avoid labels overlapping
plt.show()
```

```
In [6]: def show_average_surface_per_dept(df,department_select,types_properties_select,p
                             # We take the department, price range, slice surface and types of properties
                             selected_departments = list(department_select.value)
                             selected_types_properties = list(types_properties_select.value)
                             selected_price_range = list(price_range_select.value)
                             selected_slice_surface = list(slice_surface_select.value)
                             # If nothing is selected we take all the values for each point
                             if not selected departments:
                                      selected_departments = list(df['code_departement'].unique())
                             if not selected_types_properties:
                                      selected_types_properties = list(df['type_local'].unique())
                             if not selected price range:
                                      selected_price_range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150k €
                             if not selected slice surface:
                                      selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
                             # Filter data based with our filter
                             df filtered = df[(df['code departement'].isin(selected departments)) &
                                                                           (df['type_local'].isin(selected_types_properties))&
                                                                           (df['tranche_valeur'].isin(selected_price_range))&
                                                                           (df['plage_surface'].isin(selected_slice_surface))].copy
                             df filtered national = df[
                                                                           (df['type_local'].isin(selected_types_properties))&
                                                                           (df['tranche valeur'].isin(selected price range))&
                                                                           (df['plage_surface'].isin(selected_slice_surface))].copy
                             # We calculate the average surface area for each selected department
                             average_area_par_dept = df_filtered.groupby('code_departement')['surface_ter
                             # Calculate the average surface area of the selected departments
```

```
# Calculate the national average surface area (average of all selected depar
            national_average_area = df_filtered_national[df_filtered_national['type_local
            fig, ax = plt.subplots(figsize=(20, 6))
            # We plot the average area for each selected department
            ax.bar(average_area_par_dept.index, average_area_par_dept, label='Average Su
            # Add a bar for the average area of the selected departments
            ax.bar('Selected Average', surface_moyenne_selectionnee, label='Selected Ave
            # Add a bar for the national average area
            ax.bar('National average', national_average_area, label='National average',
            # parameters of the graphics
            ax.set_ylabel('Average surface area in m2')
            ax.set xlabel('Departments')
            ax.set_title("Average surface area in m2 by department, selection and global
            ax.grid(True, linestyle='--', alpha=0.6)
            ax.legend(title="Legend", bbox_to_anchor=(1.05, 1), loc='upper left')
            plt.xticks(rotation=90)
            plt.tight_layout()
            plt.show()
In [7]: def evolution_average_price_per_m2_per_departement_monthly(df,department_select,
            # We take the department, price range, slice surface and types of properties
            selected_departments = list(department_select.value)
            selected_types_properties = list(types_properties_select.value)
            selected_price_range = list(price_range_select.value)
            selected_slice_surface = list(slice_surface_select.value)
            # If nothing is selected we take all the values for each point
            if not selected departments:
                selected_departments = list(df['code_departement'].unique())
            if not selected_types_properties:
                selected_types_properties = list(df['type_local'].unique())
            if not selected price range:
                selected price range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150
            if not selected slice surface:
                selected slice surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
           # Filter data based with our filter
            df filtered = df[(df['code departement'].isin(selected departments)) &
                                (df['type local'].isin(selected types properties))&
                                (df['tranche_valeur'].isin(selected_price_range))&
                                (df['plage_surface'].isin(selected_slice_surface))].copy
            df_filtered_national = df[
                                (df['type local'].isin(selected types properties))&
                                (df['tranche_valeur'].isin(selected_price_range))&
                                (df['plage_surface'].isin(selected_slice_surface))].copy
            # we verify that the date column is in the correct format
            df_filtered['date_mutation'] = pd.to_datetime(df_filtered['date_mutation'])
            # we take the average price per department for each month
            average_price_par_dept_mois = df_filtered.groupby(['code_departement', 'mois']
```

surface_moyenne_selectionnee = average_area_par_dept.mean()

```
plt.figure(figsize=(15, 8))
# Add the curve for the global average
df_dept_global = df_filtered.groupby('mois_annee')['prix_m2'].mean().reset_i
plt.plot(df_dept_global['mois_annee'].astype(str), df_dept_global['prix_m2']
# Add the curve for the national average (based on the entire data)
df_national_global = df_filtered_national.groupby('mois_annee')['prix_m2'].m
plt.plot(df_national_global['mois_annee'].astype(str), df_national_global['p
# Add the curves for each selected department
for departement in selected_departments:
    df_dept = average_price_par_dept_mois[average_price_par_dept_mois['code_
    plt.plot(df_dept['mois_annee'].astype(str), df_dept['prix_m2'], marker='
plt.title("Evolution of the average price per m² by department (monthly)")
plt.xlabel("Month and year")
plt.ylabel("Average price per m² (€)")
plt.legend(title="Types of property", bbox_to_anchor=(1.05, 1), loc='upper l
plt.grid(True, linestyle='--', alpha=0.6)
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
# We take the department, price range, slice surface and types of properties
selected_departments = list(department_select.value)
selected_types_properties = list(types_properties_select.value)
```

```
In [8]: def display_number_of_room_department(df,department_select,types_properties_sele
            selected_price_range = list(price_range_select.value)
            selected_slice_surface = list(slice_surface_select.value)
            # If nothing is selected we take all the values for each point
            if not selected_departments:
                selected_departments = list(df['code_departement'].unique())
            if not selected_types_properties:
                selected_types_properties = list(df['type_local'].unique())
            if not selected_price_range:
                selected price range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150
            if not selected slice surface:
                selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
             # Filter data based with our filter
            df_filtered = df[(df['code_departement'].isin(selected_departments)) &
                                (df['type local'].isin(selected types properties))&
                                (df['tranche_valeur'].isin(selected_price_range))&
                                (df['plage_surface'].isin(selected_slice_surface))].copy
            df_filtered_national = df[
                                (df['type_local'].isin(selected_types_properties))&
                                (df['tranche_valeur'].isin(selected_price_range))&
                                (df['plage_surface'].isin(selected_slice_surface))].copy
            # We take the number of main room per department
            number_piece_per_dept = df_filtered.groupby('code_departement')['nombre_piec
            # Ajouter la moyenne générale des départements sélectionnés
            average selection = df filtered['nombre pieces principales'].mean()
            national_average = df_filtered_national['nombre_pieces_principales'].mean()
```

```
# Ajouter une barre pour la moyenne générale
departments = list(number_piece_per_dept['code_departement'])
number_room = list(number_piece_per_dept['nombre_pieces_principales'])
plt.figure(figsize=(15, 8))
plt.bar(departments, number room, color='skyblue')
#We plot line for average and national average of main rooms
plt.axhline(average_selection, color='red', linestyle='dashed', linewidth=2,
plt.axhline(national_average, color='green', linestyle='dashed', linewidth=2
plt.title("Average number of rooms per department and per type of property s
plt.xlabel("Department")
plt.ylabel("Average number of rooms")
plt.xticks(rotation=90)
plt.legend()
plt.grid(True, linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```

```
In [9]: def distribution_type_property_by_department(df,department_select,types_properti
            # We take the department, price range, slice surface and types of properties
            selected_departments = list(department_select.value)
            selected_types_properties = list(types_properties_select.value)
            selected_price_range = list(price_range_select.value)
            selected_slice_surface = list(slice_surface_select.value)
           # If nothing is selected we take all the values for each point
            if not selected_departments:
                selected_departments = list(df['code_departement'].unique())
            if not selected_types_properties:
                selected_types_properties = list(df['type_local'].unique())
            if not selected_price_range:
                selected_price_range = list(df['tranche_valeur'].unique())
            if not selected_slice_surface:
                selected_slice_surface = list(df['plage_surface'].unique())
             # Filter data based with our filter
            df_filtered = df[(df['code_departement'].isin(selected_departments)) &
                                (df['type local'].isin(selected types properties))&
                                (df['tranche_valeur'].isin(selected_price_range))&
                                (df['plage_surface'].isin(selected_slice_surface))].copy
            # Calculate the distribution of property types by department
            distribution = df_filtered.groupby(['code_departement', 'type_local']).size(
            # Normalize to obtain proportions (in %)
            distribution_norm = distribution.div(distribution.sum(axis=1), axis=0) * 100
            # creating a Stacked Bar Chart
            distribution_norm.plot(kind='bar', figsize=(15, 8), stacked=True, colormap='
            plt.title("Distribution of transactions by type of property and by departmen
            plt.xlabel("Department")
            plt.ylabel("Proportion of transactions (%)")
            plt.legend(title="Types of properties", bbox_to_anchor=(1.05, 1), loc='upper
            plt.grid(axis='y', linestyle='--', alpha=0.6)
            plt.xticks(rotation=45)
            plt.tight layout()
            plt.show()
```

```
In [10]: def distribution_of_land_value_by_tranche(df, department_select, types_propertie
              # Liste ordonnée des plages de prix
              price_order = [
                  "< 50k €", "50k - 100k €", "100k - 150k €", "150k - 200k €", "200k - 250
                  "250k - 300k \in", "300k - 350k \in", "350k - 400k \in", "400k - 450k \in", "450 "500k - 550k \in", "550k - 600k \in", "600k - 650k \in", "650k - 700k \in", "700
                  "750k - 800k €", "800k - 850k €", "850k - 900k €", "900k - 950k €", "950
                  "1M - 1.5M €", "1.5M - 2M €", "2M - 2.5M €", "2.5M - 3M €", "3M - 4M €",
              ]
              # Transformer la colonne 'tranche_valeur' en catégorie ordonnée
              df['tranche_valeur'] = pd.Categorical(df['tranche_valeur'], categories=price
              # We take the department, price range, slice surface and types of properties
              selected_departments = list(department_select.value)
              selected_types_properties = list(types_properties_select.value)
              selected_price_range = list(price_range_select.value)
              selected_slice_surface = list(slice_surface_select.value)
              # If nothing is selected we take all the values for each point
              if not selected departments:
                  selected_departments = list(df['code_departement'].unique())
              if not selected_types_properties:
                  selected_types_properties = list(df['type_local'].unique())
              if not selected_price_range:
                  selected_price_range = price_order
              if not selected_slice_surface:
                  selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
                                             '250-300 m2', '300-350 m2', '350-400 m2', '400
                                             '500-600 m2', '600-700 m2', '700-800 m2', '800
              # Filter data based with our filter
              df_filtered = df[(df['code_departement'].isin(selected_departments)) &
                                (df['type_local'].isin(selected_types_properties)) &
                               (df['tranche_valeur'].isin(selected_price_range)) &
                               (df['plage_surface'].isin(selected_slice_surface))].copy()
              # Count the number of transactions by tranche and by department
              distribution = df filtered.groupby(['code departement', 'tranche valeur']).s
              # Normalize to obtain proportions (in %)
              distribution_norm = distribution.div(distribution.sum(axis=1), axis=0) * 100
              # Create a graphic
              distribution norm.T.plot(kind='bar', figsize=(15, 8), stacked=True, colormap
              plt.title("Distribution of transactions by land value bracket and by departm
              plt.xlabel("Land value range (€)")
              plt.ylabel("Proportion of transactions (%)")
              plt.legend(title="Department", bbox_to_anchor=(1.05, 1), loc='upper left')
              plt.grid(axis='y', linestyle='--', alpha=0.6)
              plt.xticks(rotation=45)
              plt.tight layout()
              plt.show()
In [11]: def price m2 per surface(df, department select, types properties select, price ra
              # We take the department, price range, slice surface and types of properties
              selected departments = list(department select.value)
              selected_types_properties = list(types_properties_select.value)
              selected_price_range = list(price_range_select.value)
              selected_slice_surface = list(slice_surface_select.value)
```

```
# If nothing is selected we take all the values for each point
if not selected_departments:
    selected_departments = list(df['code_departement'].unique())
if not selected_types_properties:
    selected_types_properties = list(df['type_local'].unique())
if not selected price range:
    selected_price_range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150
if not selected slice surface:
    selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
# Filter data based with our filter
df_filtered = df[(df['code_departement'].isin(selected_departments)) &
                    (df['type_local'].isin(selected_types_properties))&
                    (df['tranche_valeur'].isin(selected_price_range))&
                    (df['plage_surface'].isin(selected_slice_surface))].copy
df_filtered_national = df[
                    (df['type_local'].isin(selected_types_properties))&
                    (df['tranche_valeur'].isin(selected_price_range))&
                    (df['plage_surface'].isin(selected_slice_surface))].copy
# Calculate the average price by surface area and department
average_price_per_surface = df_filtered.groupby(['plage_surface', 'code_depa
# Calculate average by surface range for selection
average_selection = df_filtered.groupby('plage_surface')['prix_m2'].mean().r
# Calculate the national average by surface area
national_average = df_filtered_national.groupby('plage_surface')['prix_m2'].
# Creating the graph
plt.figure(figsize=(12, 8))
#For each department selected, we take the average price per suface and then
for department in selected_departments:
    df department = average price per surface[average price per surface['cod
    if not df department.empty:
        plt.plot(
            df_department['plage_surface'],
            df department['prix m2'],
            marker='o',
            label=f'Department {department}'
plt.plot(
    average_selection['plage_surface'],
    average_selection['prix_m2'],
    marker='o',
    linestyle='--',
    label='Selected average',
    color='black'
# Then we add the National average line
plt.plot(
    national average['plage surface'],
    national average['prix m2'],
    marker='o',
    linestyle=':',
    label='National average',
    color='red'
)
```

```
# graph configurations
             plt.title("Average price per m² depending on surface area ranges")
             plt.xlabel("Surface range (m²)")
             plt.ylabel("Average price per m² (€)")
             plt.legend(title="Departement", bbox_to_anchor=(1.05, 1), loc='upper left')
             plt.grid(True, linestyle='--', alpha=0.6)
             plt.xticks(rotation=45)
             plt.tight_layout()
             plt.show()
In [12]: def add_region_name_column(df):
             Add a column 'region_name' to the dataframe with the name of the region corr
             :param df: DataFrame with the column 'code_insee' containing the INSEE codes
             :return: DataFrame with the added column 'region_name'
             # Dictionary of INSEE codes for regions and their names
             regions_names = {
                  "01": "Guadeloupe", "02": "Martinique", "03": "Guyane", "04": "La Réunio
                 "24": "Centre-Val de Loire", "27": "Bourgogne-Franche-Comté", "28": "Nor
                 "44": "Grand Est", "52": "Pays de la Loire", "53": "Bretagne", "75": "No
                  "84": "Auvergne-Rhône-Alpes", "93": "Provence-Alpes-Côte d'Azur", "94":
             df['region_name'] = df['code_insee'].map(regions_names)
             return df
In [13]: | def explore_france(data, department_select=None, types_properties_select=None, p
             Displays a map of France with the average price per m2 of land by region, an
             Parameters:
             - data : pd.DataFrame : The DataFrame containing the data of real estate tra
             - department_select : ipywidgets.SelectMultiple : The widget for selecting d
             - types properties select : ipywidgets.SelectMultiple : The widget for select
             - price range select : ipywidgets.SelectMultiple : The widget for selecting
             - slice_surface_select : ipywidgets.SelectMultiple : The widget for selectin
             Returns:
             - Map of France with the average price per m² of land by region, and the loc
             data_general = data.copy()
             #We take the department, price range, slice surface and types of properties s
             selected_departments = list(department_select.value)
             selected types properties = list(types properties select.value)
             selected_price_range = list(price_range_select.value)
             selected_slice_surface = list(slice_surface_select.value)
             # If nothing is selected we take all the values for each point
             if not selected_departments:
                 selected_departments = list(data['code_departement'].unique())
             if not selected_types_properties:
                 selected types properties = list(data['type local'].unique())
             if not selected_price_range:
                 selected price range = ["< 50k €", "50k - 100k €", "100k - 150k €", "150k
             if not selected_slice_surface:
```

```
selected_slice_surface = ['0-50 m2', '50-100 m2', '100-150 m2', '150-200
# Filter data based with our filter
data = data[
    (data['code_departement'].isin(selected_departments)) &
    (data['type_local'].isin(selected_types_properties)) &
    (data['tranche_valeur'].isin(selected_price_range)) &
    (data['plage_surface'].isin(selected_slice_surface))
].copy()
# Load shapefile
shp_file_path = "regions.shx"
map_france = gpd.read_file(shp_file_path)
# Define the CRS
if map_france.crs is None:
    map_france = map_france.set_crs(epsg=4326, allow_override=True)
#We verify that the column 'code insee' is in the correct format and correct
data_general["code_insee"] = data_general["code_insee"].astype(str).str.zfil
#Average price per m² by region
ratio_average = data_general.groupby('code_insee')['prix_m2'].mean()
# Add the average land value by region stylized in euros
map_france['Average_land_value'] = map_france['code_insee'].map(ratio_average)
map_france['Average_land_value'] = map_france['Average_land_value'].apply(la
# Normalize the values
ratio_average = (ratio_average - ratio_average.min()) / (ratio_average.max())
# Logarithmic transformation to accentuate differences
ratio_average = np.log1p(ratio_average * 10)
map_france = map_france.set_index('code_insee').join(ratio_average, how='lef
map france = add region name column(map france)
sample_data = data.copy() if department_select and types_properties_select a
geometry = gpd.GeoSeries.from xy(sample data['longitude'], sample data['lati
points_gdf = gpd.GeoDataFrame(sample_data, geometry=geometry)
# Creating the map
map = folium.Map(location=[46.603354, 1.888334], zoom_start=6)
map_france["prix_m2"].fillna(0, inplace=True)
# We add the regions to the map
folium.GeoJson(
    map_france,
    name='geojson',
    style_function=lambda feature: {
        'fillColor': matplotlib.colors.rgb2hex(plt.cm.plasma(feature['prope
        'color': matplotlib.colors.rgb2hex(plt.cm.plasma(feature['properties')
        'weight': 2,
        'fillOpacity': 0.5,
    },
    tooltip=folium.GeoJsonTooltip(fields=['region_name', 'Average_land_value
).add_to(map)
```

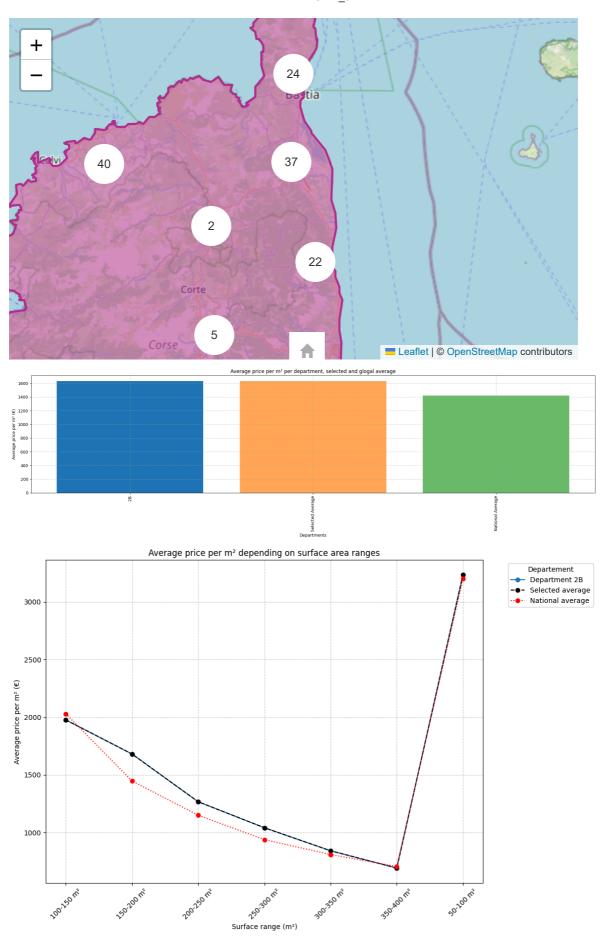
```
# Add some points
marker_cluster = MarkerCluster().add_to(map)

for idx, row in points_gdf.iterrows():
    folium.Marker(
        location=[row['latitude'], row['longitude']],
        popup=f"{row.get('region_name', 'N/A')} \n Type : {row.get('type_loc_icon=folium.Icon(color='red', icon='home'))
    ).add_to(marker_cluster)

display(map)
    return map
```

III. Exécution des fonctions

```
In [16]: selection filters and analyses(data)
        SelectMultiple(description='Departments', layout=Layout(width='60%'), options=('0
        1', '02', '03', '04', '05', '...
        SelectMultiple(description='Types of property', layout=Layout(width='60%'), optio
        ns=('House', 'Dependency (Iso...
        SelectMultiple(description='Price range', layout=Layout(width='60%'), options=('<</pre>
        50k €', '50k - 100k €', '100...
        SelectMultiple(description='Surface slice', layout=Layout(width='60%'), options=
        ('0-50 m<sup>2</sup>', '50-100 m<sup>2</sup>', '100-...
        SelectMultiple(description='Analyse', layout=Layout(width='60%'), options=('Avera
        ge price per m²', 'Average su...
        Button(button_style='success', description='Apply the filters', style=ButtonStyle
        ())
        C:\Users\Arthur\AppData\Local\Temp\ipykernel_25200\3141295893.py:76: FutureWarnin
        g: A value is trying to be set on a copy of a DataFrame or Series through chained
        assignment using an inplace method.
        The behavior will change in pandas 3.0. This inplace method will never work becau
        se the intermediate object on which we are setting values always behaves as a cop
        у.
        For example, when doing 'df[col].method(value, inplace=True)', try using 'df.meth
        od({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to pe
        rform the operation inplace on the original object.
          map_france["prix_m2"].fillna(0, inplace=True)
```





The investor is looking for a property in North Corsica, close to the sea (department code: 2B) to generate location rands, he has a maximum budget of €450,000 but he does not know what type of property he can find for his criteria: In the first graph (interactive map) we can see a distribution of properties with our criteria, there are 153, most of the properties are close to the sea because in Corsica most of the cities are on the coast. There are more properties on the east coast and cheaper.

In the second graph we can see the average price per m2 in North Corsica compared to the national price and North Corsica is much more expensive certainly because it is a

beautiful destination and a seaside region.

In the third graph we can see the price ratio m2 by size of real estate in North Corsica, it decreases sharply, it is divided by 2-3 between 100m2 and 400m2. The investor must therefore buy a large surface area but it could be useless for rental.

In the 5th graph we can see that the average price is lower than the national average since 2024, which may be the best time to invest and buy real estate.

In the 6th graph we can see the distribution of property types, in North Corsica there are only a few apartments and many houses, so a house could be the best choice because: There are many houses so a lot of choice and tenants will prefer the house for privacy, space and garden.

In conclusion and according to our analysis, the best places to invest in real estate for our investor are Bastia (East coast) or Calvi (West coast) for the choice of the house, Calvi should be more expensive to buy but the investor should earn more profits with the location. The real estate should be a house because it is the most popular and this house should have a garden if it is pleasant or will be easily because the price per m2 will be lower. The price of real estate depends on the city between Bastia and Calvi but it is between 200,000€ and 450,000€