

```
In [2]: pip install geopandas
```

```
Collecting geopandas
  Obtaining dependency information for geopandas from https://files.pytho
sted.org/packages/c4/64/7d344cfcef5efddf9cf32f59af7f855828e9d74b5f862eddf5bf
d9f25323/geopandas-1.0.1-py3-none-any.whl.metadata
  Downloading geopandas-1.0.1-py3-none-any.whl.metadata (2.2 kB)
Requirement already satisfied: numpy>=1.22 in ./anaconda3/lib/python3.11/sit
e-packages (from geopandas) (1.24.3)
Collecting pyogrio>=0.7.2 (from geopandas)
  Obtaining dependency information for pyogrio>=0.7.2 from https://files.pyt
honhosted.org/packages/8d/2c/c761e6adeb81bd4029a137b3240e7214a8c9aaf22588335
6196afd6ef9d8/pyogrio-0.10.0-cp311-cp311-macosx_12_0_arm64.whl.metadata
  Downloading pyogrio-0.10.0-cp311-cp311-macosx_12_0_arm64.whl.metadata (5.5
kB)
Requirement already satisfied: packaging in ./anaconda3/lib/python3.11/site-
packages (from geopandas) (23.0)
Requirement already satisfied: pandas>=1.4.0 in ./anaconda3/lib/python3.11/s
ite-packages (from geopandas) (1.5.3)
Collecting pyproj>=3.3.0 (from geopandas)
  Obtaining dependency information for pyproj>=3.3.0 from https://files.pyth
onhosted.org/packages/2d/4d/610fe2a17de71b4fe210af69ce25f2d65379ba0a48299129
894d0d0988ee/pyproj-3.7.0-cp311-cp311-macosx_14_0_arm64.whl.metadata
  Downloading pyproj-3.7.0-cp311-cp311-macosx_14_0_arm64.whl.metadata (31 k
B)
Collecting shapely>=2.0.0 (from geopandas)
  Obtaining dependency information for shapely>=2.0.0 from https://files.pyt
honhosted.org/packages/37/63/e182e43081fffa0a2d970c480f2ef91647a6ab94098f617
48c23c2a485f2/shapely-2.0.6-cp311-cp311-macosx_11_0_arm64.whl.metadata
  Downloading shapely-2.0.6-cp311-cp311-macosx_11_0_arm64.whl.metadata (7.0
kB)
Requirement already satisfied: python-dateutil>=2.8.1 in ./anaconda3/lib/pyt
hon3.11/site-packages (from pandas>=1.4.0->geopandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/si
te-packages (from pandas>=1.4.0->geopandas) (2022.7)
Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-pa
ckages (from pyogrio>=0.7.2->geopandas) (2023.7.22)
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-p
ackages (from python-dateutil>=2.8.1->pandas>=1.4.0->geopandas) (1.16.0)
Downloading geopandas-1.0.1-py3-none-any.whl (323 kB)
_____ 323.6/323.6 kB 685.1 kB/s eta 0:
00:00a 0:00:01
Downloading pyogrio-0.10.0-cp311-cp311-macosx_12_0_arm64.whl (15.1 MB)
_____ 15.1/15.1 MB 2.1 MB/s eta 0:00:
0000:0100:01m
Downloading pyproj-3.7.0-cp311-cp311-macosx_14_0_arm64.whl (4.6 MB)
_____ 4.6/4.6 MB 7.9 MB/s eta 0:00:00
00:0100:01
Downloading shapely-2.0.6-cp311-cp311-macosx_11_0_arm64.whl (1.3 MB)
_____ 1.3/1.3 MB 10.2 MB/s eta 0:00:0
0a 0:00:01
Installing collected packages: shapely, pyproj, pyogrio, geopandas
Successfully installed geopandas-1.0.1 pyogrio-0.10.0 pyproj-3.7.0 shapely-
2.0.6
Note: you may need to restart the kernel to use updated packages.
```

```
In [3]: pip install fiona shapely pyproj rtree
```

Collecting fiona

Obtaining dependency information for fiona from [https://files.pythonhosted.org/packages/65/0c/e8070b15c8303f60bd4444a120842597ccd6ed550548948e2e36cffbaa93/fiona-1.10.1-cp311-cp311-macosx\\_11\\_0\\_arm64.whl.metadata](https://files.pythonhosted.org/packages/65/0c/e8070b15c8303f60bd4444a120842597ccd6ed550548948e2e36cffbaa93/fiona-1.10.1-cp311-cp311-macosx_11_0_arm64.whl.metadata)

Downloading fiona-1.10.1-cp311-cp311-macosx\_11\_0\_arm64.whl.metadata (56 kB)

56.6/56.6 kB 1.5 MB/s eta 0:00:00a 0:00:01

Requirement already satisfied: shapely in ./anaconda3/lib/python3.11/site-packages (2.0.6)

Requirement already satisfied: pyproj in ./anaconda3/lib/python3.11/site-packages (3.7.0)

Requirement already satisfied: rtree in ./anaconda3/lib/python3.11/site-packages (1.0.1)

Requirement already satisfied: attrs>=19.2.0 in ./anaconda3/lib/python3.11/site-packages (from fiona) (23.2.0)

Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from fiona) (2023.7.22)

Requirement already satisfied: click~=8.0 in ./anaconda3/lib/python3.11/site-packages (from fiona) (8.0.4)

Collecting click-plugins>=1.0 (from fiona)

Obtaining dependency information for click-plugins>=1.0 from [https://files.pythonhosted.org/packages/e9/da/824b92d9942f4e472702488857914bdd50f73021efea15b4cad9aca8ecef/click\\_plugins-1.1.1-py2.py3-none-any.whl.metadata](https://files.pythonhosted.org/packages/e9/da/824b92d9942f4e472702488857914bdd50f73021efea15b4cad9aca8ecef/click_plugins-1.1.1-py2.py3-none-any.whl.metadata)

Downloading click\_plugins-1.1.1-py2.py3-none-any.whl.metadata (6.4 kB)

Collecting cligj>=0.5 (from fiona)

Obtaining dependency information for cligj>=0.5 from <https://files.pythonhosted.org/packages/73/86/43fa9f15c5b9fb6e82620428827cd3c284aa933431405d1bcf5231ae3d3e/cligj-0.7.2-py3-none-any.whl.metadata>

Downloading cligj-0.7.2-py3-none-any.whl.metadata (5.0 kB)

Requirement already satisfied: numpy<3,>=1.14 in ./anaconda3/lib/python3.11/site-packages (from shapely) (1.24.3)

Downloading fiona-1.10.1-cp311-cp311-macosx\_11\_0\_arm64.whl (14.8 MB)

14.8/14.8 MB 4.1 MB/s eta 0:00:0000:0100:01

Downloading click\_plugins-1.1.1-py2.py3-none-any.whl (7.5 kB)

Downloading cligj-0.7.2-py3-none-any.whl (7.1 kB)

Installing collected packages: cligj, click-plugins, fiona

Successfully installed click-plugins-1.1.1 cligj-0.7.2 fiona-1.10.1

Note: you may need to restart the kernel to use updated packages.

```
In [5]: pip install geopandas shapely
```

Requirement already satisfied: geopandas in ./anaconda3/lib/python3.11/site-packages (1.0.1)  
Requirement already satisfied: shapely in ./anaconda3/lib/python3.11/site-packages (2.0.6)  
Requirement already satisfied: numpy>=1.22 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (1.24.3)  
Requirement already satisfied: pyogrio>=0.7.2 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (0.10.0)  
Requirement already satisfied: packaging in ./anaconda3/lib/python3.11/site-packages (from geopandas) (23.0)  
Requirement already satisfied: pandas>=1.4.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (1.5.3)  
Requirement already satisfied: pyproj>=3.3.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (3.7.0)  
Requirement already satisfied: python-dateutil>=2.8.1 in ./anaconda3/lib/python3.11/site-packages (from pandas>=1.4.0->geopandas) (2.8.2)  
Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/site-packages (from pandas>=1.4.0->geopandas) (2022.7)  
Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from pyogrio>=0.7.2->geopandas) (2023.7.22)  
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.8.1->pandas>=1.4.0->geopandas) (1.16.0)  
Note: you may need to restart the kernel to use updated packages.

In [14]: `pip install rasterio numpy geopandas`

```

Collecting rasterio
  Obtaining dependency information for rasterio from https://files.pythonhosted.org/packages/b3/59/ca86697161206233eea6353237b0c0f02f6f70434144db162f964a7e1b19/rasterio-1.4.3-cp311-cp311-macosx_14_0_arm64.whl.metadata
  Downloading rasterio-1.4.3-cp311-cp311-macosx_14_0_arm64.whl.metadata (9.1 kB)
Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (1.24.3)
Requirement already satisfied: geopandas in ./anaconda3/lib/python3.11/site-packages (1.0.1)
Collecting affine (from rasterio)
  Obtaining dependency information for affine from https://files.pythonhosted.org/packages/0b/f7/85273299ab57117850cc0a936c64151171fac4da49bc6fba0dad984a7c5f/affine-2.4.0-py3-none-any.whl.metadata
  Downloading affine-2.4.0-py3-none-any.whl.metadata (4.0 kB)
Requirement already satisfied: attrs in ./anaconda3/lib/python3.11/site-packages (from rasterio) (23.2.0)
Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2023.7.22)
Requirement already satisfied: click>=4.0 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (8.0.4)
Requirement already satisfied: cligj>=0.5 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (0.7.2)
Requirement already satisfied: click-plugins in ./anaconda3/lib/python3.11/site-packages (from rasterio) (1.1.1)
Requirement already satisfied: pyparsing in ./anaconda3/lib/python3.11/site-packages (from rasterio) (3.0.9)
Requirement already satisfied: pyogrio>=0.7.2 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (0.10.0)
Requirement already satisfied: packaging in ./anaconda3/lib/python3.11/site-packages (from geopandas) (23.0)
Requirement already satisfied: pandas>=1.4.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (1.5.3)
Requirement already satisfied: pyproj>=3.3.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (3.7.0)
Requirement already satisfied: shapely>=2.0.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (2.0.6)
Requirement already satisfied: python-dateutil>=2.8.1 in ./anaconda3/lib/python3.11/site-packages (from pandas>=1.4.0->geopandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/site-packages (from pandas>=1.4.0->geopandas) (2022.7)
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.8.1->pandas>=1.4.0->geopandas) (1.16.0)
Downloading rasterio-1.4.3-cp311-cp311-macosx_14_0_arm64.whl (18.8 MB)
18.8/18.8 MB 317.6 kB/s eta 0:0
0:0000:0100:02
Downloading affine-2.4.0-py3-none-any.whl (15 kB)
Installing collected packages: affine, rasterio
Successfully installed affine-2.4.0 rasterio-1.4.3
Note: you may need to restart the kernel to use updated packages.

```

```
In [18]: pip install numpy rasterio
```

Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (1.24.3)  
 Requirement already satisfied: rasterio in ./anaconda3/lib/python3.11/site-packages (1.4.3)  
 Requirement already satisfied: affine in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2.4.0)  
 Requirement already satisfied: attrs in ./anaconda3/lib/python3.11/site-packages (from rasterio) (23.2.0)  
 Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2023.7.22)  
 Requirement already satisfied: click>=4.0 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (8.0.4)  
 Requirement already satisfied: cligj>=0.5 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (0.7.2)  
 Requirement already satisfied: click-plugins in ./anaconda3/lib/python3.11/site-packages (from rasterio) (1.1.1)  
 Requirement already satisfied: pyparsing in ./anaconda3/lib/python3.11/site-packages (from rasterio) (3.0.9)  
 Note: you may need to restart the kernel to use updated packages.

In [43]: `pip install folium`

```
Collecting folium
  Obtaining dependency information for folium from https://files.pythonhosted.org/packages/fc/ab/d1f47c48a14e17cd487c8b467b573291fae75477b067241407e7889a3692/folium-0.19.4-py2.py3-none-any.whl.metadata
  Downloading folium-0.19.4-py2.py3-none-any.whl.metadata (3.8 kB)
Collecting branca>=0.6.0 (from folium)
  Obtaining dependency information for branca>=0.6.0 from https://files.pythonhosted.org/packages/f8/9d/91cddd38bd00170aad1a4b198c47b4ed716be45c234e09b835af41f4e717/branca-0.8.1-py3-none-any.whl.metadata
  Downloading branca-0.8.1-py3-none-any.whl.metadata (1.5 kB)
Requirement already satisfied: Jinja2>=2.9 in ./anaconda3/lib/python3.11/site-packages (from folium) (3.1.2)
Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (from folium) (1.24.3)
Requirement already satisfied: requests in ./anaconda3/lib/python3.11/site-packages (from folium) (2.31.0)
Requirement already satisfied: xyzservices in ./anaconda3/lib/python3.11/site-packages (from folium) (2022.9.0)
Requirement already satisfied: MarkupSafe>=2.0 in ./anaconda3/lib/python3.11/site-packages (from Jinja2>=2.9->folium) (2.1.1)
Requirement already satisfied: charset-normalizer<4,>=2 in ./anaconda3/lib/python3.11/site-packages (from requests->folium) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in ./anaconda3/lib/python3.11/site-packages (from requests->folium) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in ./anaconda3/lib/python3.11/site-packages (from requests->folium) (1.26.16)
Requirement already satisfied: certifi>=2017.4.17 in ./anaconda3/lib/python3.11/site-packages (from requests->folium) (2023.7.22)
Downloading folium-0.19.4-py2.py3-none-any.whl (110 kB)
110.5/110.5 kB 924.8 kB/s eta 0:00:00a 0:00:01
Downloading branca-0.8.1-py3-none-any.whl (26 kB)
Installing collected packages: branca, folium
Successfully installed branca-0.8.1 folium-0.19.4
Note: you may need to restart the kernel to use updated packages.
```

```
In [50]: pip install numpy pandas
```

```
Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (1.24.3)
Requirement already satisfied: pandas in ./anaconda3/lib/python3.11/site-packages (1.5.3)
Requirement already satisfied: python-dateutil>=2.8.1 in ./anaconda3/lib/python3.11/site-packages (from pandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/site-packages (from pandas) (2022.7)
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.8.1->pandas) (1.16.0)
Note: you may need to restart the kernel to use updated packages.
```

```
In [62]: pip install numpy rasterio matplotlib
```

```
Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (1.24.3)
Requirement already satisfied: rasterio in ./anaconda3/lib/python3.11/site-packages (1.4.3)
Requirement already satisfied: matplotlib in ./anaconda3/lib/python3.11/site-packages (3.7.1)
Requirement already satisfied: affine in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2.4.0)
Requirement already satisfied: attrs in ./anaconda3/lib/python3.11/site-packages (from rasterio) (23.2.0)
Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2023.7.22)
Requirement already satisfied: click>=4.0 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (8.0.4)
Requirement already satisfied: cligj>=0.5 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (0.7.2)
Requirement already satisfied: click-plugins in ./anaconda3/lib/python3.11/site-packages (from rasterio) (1.1.1)
Requirement already satisfied: pyparsing in ./anaconda3/lib/python3.11/site-packages (from rasterio) (3.0.9)
Requirement already satisfied: contourpy>=1.0.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (1.0.5)
Requirement already satisfied: cycycler>=0.10 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (4.25.0)
Requirement already satisfied: kiwisolver>=1.0.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (1.4.4)
Requirement already satisfied: packaging>=20.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (23.0)
Requirement already satisfied: pillow>=6.2.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (9.4.0)
Requirement already satisfied: python-dateutil>=2.7 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)
Note: you may need to restart the kernel to use updated packages.
```

```
In [68]: !pip install jenkspy
```

Requirement already satisfied: jenkspy in ./anaconda3/lib/python3.11/site-packages (0.4.1)

Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/site-packages (from jenkspy) (1.24.3)

In [77]: `pip install geopandas shapely matplotlib`

Requirement already satisfied: geopandas in ./anaconda3/lib/python3.11/site-packages (1.0.1)

Requirement already satisfied: shapely in ./anaconda3/lib/python3.11/site-packages (2.0.6)

Requirement already satisfied: matplotlib in ./anaconda3/lib/python3.11/site-packages (3.7.1)

Requirement already satisfied: numpy>=1.22 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (1.24.3)

Requirement already satisfied: pyogrio>=0.7.2 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (0.10.0)

Requirement already satisfied: packaging in ./anaconda3/lib/python3.11/site-packages (from geopandas) (23.0)

Requirement already satisfied: pandas>=1.4.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (1.5.3)

Requirement already satisfied: pyproj>=3.3.0 in ./anaconda3/lib/python3.11/site-packages (from geopandas) (3.7.0)

Requirement already satisfied: contourpy>=1.0.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (1.0.5)

Requirement already satisfied: cyclor>=0.10 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (0.11.0)

Requirement already satisfied: fonttools>=4.22.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (4.25.0)

Requirement already satisfied: kiwisolver>=1.0.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (1.4.4)

Requirement already satisfied: pillow>=6.2.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (9.4.0)

Requirement already satisfied: pyparsing>=2.3.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (3.0.9)

Requirement already satisfied: python-dateutil>=2.7 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/site-packages (from pandas>=1.4.0->geopandas) (2022.7)

Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from pyogrio>=0.7.2->geopandas) (2023.7.22)

Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

In [84]: `pip install streamlit geopandas matplotlib rasterio`

Collecting streamlit

Obtaining dependency information for streamlit from <https://files.pythonhosted.org/packages/c2/87/b2e162869500062a94dde7589c167367b5538dab6eacce2e7c0f00d5c9c5/streamlit-1.41.1-py2.py3-none-any.whl.metadata>

Downloading streamlit-1.41.1-py2.py3-none-any.whl.metadata (8.5 kB)

Requirement already satisfied: geopandas in ./anaconda3/lib/python3.11/site-packages (1.0.1)

Requirement already satisfied: matplotlib in ./anaconda3/lib/python3.11/site-packages (3.7.1)

Requirement already satisfied: rasterio in ./anaconda3/lib/python3.11/site-p



```
ackages (1.4.3)
Collecting altair<6,>=4.0 (from streamlit)
  Obtaining dependency information for altair<6,>=4.0 from https://files.pyth
onhosted.org/packages/aa/f3/0b6ced594e51cc95d8c1fc1640d3623770d01e4969d29c0
bd09945fafefa/altair-5.5.0-py3-none-any.whl.metadata
  Downloading altair-5.5.0-py3-none-any.whl.metadata (11 kB)
Collecting blinker<2,>=1.0.0 (from streamlit)
  Obtaining dependency information for blinker<2,>=1.0.0 from https://files.
pythonhosted.org/packages/10/cb/f2ad4230dc2eb1a74edf38f1a38b9b52277f75bef262
d8908e60d957e13c/blinker-1.9.0-py3-none-any.whl.metadata
  Downloading blinker-1.9.0-py3-none-any.whl.metadata (1.6 kB)
Collecting cachetools<6,>=4.0 (from streamlit)
  Obtaining dependency information for cachetools<6,>=4.0 from https://file
s.pythonhosted.org/packages/a4/07/14f8ad37f2d12a5ce41206c21820d8cb6561b728e5
1fad4530dff0552a67/cachetools-5.5.0-py3-none-any.whl.metadata
  Downloading cachetools-5.5.0-py3-none-any.whl.metadata (5.3 kB)
Requirement already satisfied: click<9,>=7.0 in ./anaconda3/lib/python3.11/s
ite-packages (from streamlit) (8.0.4)
Requirement already satisfied: numpy<3,>=1.23 in ./anaconda3/lib/python3.11/
site-packages (from streamlit) (1.24.3)
Requirement already satisfied: packaging<25,>=20 in ./anaconda3/lib/python3.
11/site-packages (from streamlit) (23.0)
Requirement already satisfied: pandas<3,>=1.4.0 in ./anaconda3/lib/python3.1
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Requirement already satisfied: pillow<12,>=7.1.0 in ./anaconda3/lib/python3.
11/site-packages (from streamlit) (9.4.0)
Requirement already satisfied: protobuf<6,>=3.20 in ./anaconda3/lib/python3.
11/site-packages (from streamlit) (4.25.4)
Requirement already satisfied: pyarrow>=7.0 in ./anaconda3/lib/python3.11/si
te-packages (from streamlit) (11.0.0)
Requirement already satisfied: requests<3,>=2.27 in ./anaconda3/lib/python3.
11/site-packages (from streamlit) (2.31.0)
Requirement already satisfied: rich<14,>=10.14.0 in ./anaconda3/lib/python3.
11/site-packages (from streamlit) (13.8.0)
Requirement already satisfied: tenacity<10,>=8.1.0 in ./anaconda3/lib/python
3.11/site-packages (from streamlit) (8.2.2)
Requirement already satisfied: toml<2,>=0.10.1 in ./anaconda3/lib/python3.1
1/site-packages (from streamlit) (0.10.2)
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b/python3.11/site-packages (from streamlit) (4.11.0)
Collecting gitpython!=3.1.19,<4,>=3.0.7 (from streamlit)
  Obtaining dependency information for gitpython!=3.1.19,<4,>=3.0.7 from htt
ps://files.pythonhosted.org/packages/1d/9a/4114a9057db2f1462d5c8f8390ab73839
25felac012eaa42402ad65c2963/GitPython-3.1.44-py3-none-any.whl.metadata
  Downloading GitPython-3.1.44-py3-none-any.whl.metadata (13 kB)
Collecting pydeck<1,>=0.8.0b4 (from streamlit)
  Obtaining dependency information for pydeck<1,>=0.8.0b4 from https://file
s.pythonhosted.org/packages/ab/4c/b888e6cf58bd9db9c93f40d1c6be8283ff49d88919
231afe93a6bcf61626/pydeck-0.9.1-py2.py3-none-any.whl.metadata
  Downloading pydeck-0.9.1-py2.py3-none-any.whl.metadata (4.1 kB)
Requirement already satisfied: tornado<7,>=6.0.3 in ./anaconda3/lib/python3.
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ite-packages (from geopandas) (3.7.0)
Requirement already satisfied: shapely>=2.0.0 in ./anaconda3/lib/python3.11/
```



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site-packages (from geopandas) (2.0.6)
Requirement already satisfied: contourpy>=1.0.1 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (1.0.5)
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Requirement already satisfied: fonttools>=4.22.0 in ./anaconda3/lib/python3.11/site-packages (from matplotlib) (4.25.0)
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Requirement already satisfied: affine in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2.4.0)
Requirement already satisfied: attrs in ./anaconda3/lib/python3.11/site-packages (from rasterio) (23.2.0)
Requirement already satisfied: certifi in ./anaconda3/lib/python3.11/site-packages (from rasterio) (2023.7.22)
Requirement already satisfied: cligj>=0.5 in ./anaconda3/lib/python3.11/site-packages (from rasterio) (0.7.2)
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Requirement already satisfied: Jinja2 in ./anaconda3/lib/python3.11/site-packages (from altair<6,>=4.0->streamlit) (3.1.2)
Requirement already satisfied: jsonschema>=3.0 in ./anaconda3/lib/python3.11/site-packages (from altair<6,>=4.0->streamlit) (4.17.3)
Collecting narwhals>=1.14.2 (from altair<6,>=4.0->streamlit)
  Obtaining dependency information for narwhals>=1.14.2 from https://files.pythonhosted.org/packages/17/d9/00a937201a3ca6e8b2ca29226935be53dd7bc9bee27cb8ba5b27efellccf/narwhals-1.22.0-py3-none-any.whl.metadata
  Downloading narwhals-1.22.0-py3-none-any.whl.metadata (10.0 kB)
Collecting gitdb<5,>=4.0.1 (from gitpython!=3.1.19,<4,>=3.0.7->streamlit)
  Obtaining dependency information for gitdb<5,>=4.0.1 from https://files.pythonhosted.org/packages/a0/61/5c78b91c3143ed5c14207f463aecfc8f9dbb5092fb2869baf37c273b2705/gitdb-4.0.12-py3-none-any.whl.metadata
  Downloading gitdb-4.0.12-py3-none-any.whl.metadata (1.2 kB)
Requirement already satisfied: pytz>=2020.1 in ./anaconda3/lib/python3.11/site-packages (from pandas<3,>=1.4.0->streamlit) (2022.7)
Requirement already satisfied: six>=1.5 in ./anaconda3/lib/python3.11/site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)
Requirement already satisfied: charset-normalizer<4,>=2 in ./anaconda3/lib/python3.11/site-packages (from requests<3,>=2.27->streamlit) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in ./anaconda3/lib/python3.11/site-packages (from requests<3,>=2.27->streamlit) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in ./anaconda3/lib/python3.11/site-packages (from requests<3,>=2.27->streamlit) (1.26.16)
Requirement already satisfied: markdown-it-py>=2.2.0 in ./anaconda3/lib/python3.11/site-packages (from rich<14,>=10.14.0->streamlit) (2.2.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in ./anaconda3/lib/python3.11/site-packages (from rich<14,>=10.14.0->streamlit) (2.15.1)
Collecting smmap<6,>=3.0.1 (from gitdb<5,>=4.0.1->gitpython!=3.1.19,<4,>=3.0.7->streamlit)
  Obtaining dependency information for smmap<6,>=3.0.1 from https://files.pythonhosted.org/packages/04/be/d09147ad1ec7934636ad912901c5fd7667e1c858e19d355237db0d0cd5e4/smmap-5.0.2-py3-none-any.whl.metadata
  Downloading smmap-5.0.2-py3-none-any.whl.metadata (4.3 kB)

```

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Requirement already satisfied: MarkupSafe>=2.0 in ./anaconda3/lib/python3.11/site-packages (from jinja2->altair<6,>=4.0->streamlit) (2.1.1)
Requirement already satisfied: pyrsistent!=0.17.0,!0.17.1,!0.17.2,>=0.14.0 in ./anaconda3/lib/python3.11/site-packages (from jsonschema>=3.0->altair<6,>=4.0->streamlit) (0.18.0)
Requirement already satisfied: mdurl~=0.1 in ./anaconda3/lib/python3.11/site-packages (from markdown-it-py>=2.2.0->rich<14,>=10.14.0->streamlit) (0.1.0)
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Downloading blinker-1.9.0-py3-none-any.whl (8.5 kB)
Downloading cachetools-5.5.0-py3-none-any.whl (9.5 kB)
Downloading GitPython-3.1.44-py3-none-any.whl (207 kB)
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Downloading gitdb-4.0.12-py3-none-any.whl (62 kB)
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Downloading narwhals-1.22.0-py3-none-any.whl (297 kB)
_____ 297.5/297.5 kB 15.2 MB/s eta 0:00:00
00:00
Downloading smmap-5.0.2-py3-none-any.whl (24 kB)
Installing collected packages: smmap, narwhals, cachetools, blinker, pydeck, gitdb, altair, gitpython, streamlit
Successfully installed altair-5.5.0 blinker-1.9.0 cachetools-5.5.0 gitdb-4.0.12 gitpython-3.1.44 narwhals-1.22.0 pydeck-0.9.1 smmap-5.0.2 streamlit-1.41.1
Note: you may need to restart the kernel to use updated packages.

```

```

In [85]: import geopandas as gpd
import pandas as pd
from shapely.geometry import Point, LineString, Polygon
import numpy as np
import rasterio
from rasterio.transform import from_origin
import os
from rasterio.warp import calculate_default_transform, reproject, Resampling
import matplotlib.pyplot as plt
import folium
from rasterio.plot import reshape_as_image
import seaborn as sns
from matplotlib import patches
import jenks
import streamlit as st

```

```
In [24]: # Grid properties
width, height = 100, 100
cell_size = 0.01
x_min, y_max = 29.0, 39.8 # Top-left corner of the grid

# Synthetic raster data
lkp_layer = np.random.rand(height, width) # LKP likelihood layer

# Transform and save the raster
transform = from_origin(x_min, y_max, cell_size, cell_size)

with rasterio.open(
    "synthetic_lkp_raster.tif",
    "w",
    driver="GTiff",
    height=height,
    width=width,
    count=1,
    dtype=lkp_layer.dtype,
    crs="EPSG:4326",
    transform=transform,
) as dst:
    dst.write(lkp_layer, 1)

print("Synthetic raster saved as 'synthetic_lkp_raster.tif'")
```

Synthetic raster saved as 'synthetic\_lkp\_raster.tif'

```

In [13]: # Ensure the directory exists
output_dir = "./output_data"
os.makedirs(output_dir, exist_ok=True)

# Step 1: Create settlement points (villages)
settlements_data = {
    "name": ["Village A", "Village B", "Village C"],
    "geometry": [Point(29.3, 39.4), Point(29.4, 39.5), Point(29.5, 39.3)],
}
settlements = gpd.GeoDataFrame(settlements_data, crs="EPSG:4326") # WGS84 C

# Step 2: Reproject to a projected CRS for buffer calculation (e.g., UTM Zone 32N)
settlements = settlements.to_crs(epsg=32635)

# Step 3: Add buffers around settlements (e.g., 5 km buffer)
settlements["buffer"] = settlements.geometry.buffer(5000) # Buffer in meters

# Create a separate GeoDataFrame for buffers
settlements_buffers = settlements.copy()
settlements_buffers = settlements_buffers.set_geometry("buffer").drop(columns=["geometry"])

# Step 4: Create road lines in the same CRS
roads_data = {
    "name": ["Road 1", "Road 2"],
    "geometry": [
        LineString([(629000, 4350000), (635000, 4390000)]),
        LineString([(631000, 4350000), (631000, 4390000)]),
    ],
}
roads = gpd.GeoDataFrame(roads_data, crs="EPSG:32635")

# Step 5: Create a region of interest (polygon)
region_polygon = Polygon([(628000, 4340000), (637000, 4340000), (637000, 4400000), (628000, 4400000)])
region = gpd.GeoDataFrame({"name": ["Region of Interest"], "geometry": [region_polygon]})

# Step 6: Save the data to files
settlements.to_crs(epsg=4326).drop(columns=["buffer"]).to_file(f"{output_dir}/settlements.shp")
settlements_buffers.to_crs(epsg=4326).to_file(f"{output_dir}/settlements_buffers.shp")
roads.to_crs(epsg=4326).to_file(f"{output_dir}/roads.shp") # Save roads
region.to_crs(epsg=4326).to_file(f"{output_dir}/region.shp") # Save region

f"Shapefiles saved successfully in {output_dir}."

```

```

Out[13]: 'Shapefiles saved successfully in ./output_data.'

```

```

In [17]: # Step 1: Define grid properties
width, height = 100, 100 # Grid size
cell_size = 0.01 # Cell size in degrees
x_min, y_max = 29.0, 39.8 # Top-left corner of the grid

# Step 2: Create raster layers for criteria
# Example: LKP influence (high probability near center)
x_center, y_center = 29.4, 39.5
x = np.linspace(x_min, x_min + width * cell_size, width)
y = np.linspace(y_max, y_max - height * cell_size, height)
xv, yv = np.meshgrid(x, y)

# Compute distance from LKP
lkp_layer = np.exp(-np.sqrt((xv - x_center)**2 + (yv - y_center)**2) / 0.05)

# Example: Settlement influence (higher probability in certain areas)
settlement_layer = np.zeros((height, width))
settlement_coords = [(29.3, 39.4), (29.5, 39.3)]
for sx, sy in settlement_coords:
    settlement_layer += np.exp(-np.sqrt((xv - sx)**2 + (yv - sy)**2) / 0.03)

# Example: Terrain influence (random example values)
terrain_layer = np.random.rand(height, width)

# Step 3: Weighted overlay
weights = {
    "lkp": 0.5,
    "settlement": 0.3,
    "terrain": 0.2,
}
combined_layer = (weights["lkp"] * lkp_layer +
                  weights["settlement"] * settlement_layer +
                  weights["terrain"] * terrain_layer)

# Step 4: Save the probability map as a GeoTIFF
transform = from_origin(x_min, y_max, cell_size, cell_size)
with rasterio.open(
    "probability_map.tif",
    "w",
    driver="GTiff",
    height=combined_layer.shape[0],
    width=combined_layer.shape[1],
    count=1,
    dtype=combined_layer.dtype,
    crs="EPSG:4326",
    transform=transform,
) as dst:
    dst.write(combined_layer, 1)

print("Probability map saved as 'probability_map.tif'")

```

Probability map saved as 'probability\_map.tif'

```

In [19]: # Step 1: Define grid properties
width, height = 100, 100 # Grid dimensions
cell_size = 0.01 # Grid resolution in degrees

```

```

x_min, y_max = 29.0, 39.8 # Top-left corner of the grid

# Step 2: Create synthetic layers for criteria
# LKP influence (higher near the center)
x_center, y_center = 29.4, 39.5
x = np.linspace(x_min, x_min + width * cell_size, width)
y = np.linspace(y_max, y_max - height * cell_size, height)
xv, yv = np.meshgrid(x, y)

lkp_layer = np.exp(-np.sqrt((xv - x_center)**2 + (yv - y_center)**2) / 0.05)

# Settlement influence (higher near predefined points)
settlement_layer = np.zeros((height, width))
settlement_coords = [(29.3, 39.4), (29.5, 39.3)]
for sx, sy in settlement_coords:
    settlement_layer += np.exp(-np.sqrt((xv - sx)**2 + (yv - sy)**2) / 0.03)

# Terrain influence (random example values)
terrain_layer = np.random.rand(height, width)

# Step 3: Normalize layers to [0, 1]
lkp_norm = (lkp_layer - lkp_layer.min()) / (lkp_layer.max() - lkp_layer.min())
settlement_norm = (settlement_layer - settlement_layer.min()) / (settlement_layer.max() - settlement_layer.min())
terrain_norm = (terrain_layer - terrain_layer.min()) / (terrain_layer.max() - terrain_layer.min())

# Step 4: Assign weights to criteria
weights = {
    "lkp": 0.5,
    "settlement": 0.3,
    "terrain": 0.2,
}

# Step 5: Compute the weighted sum
saw_result = (
    weights["lkp"] * lkp_norm +
    weights["settlement"] * settlement_norm +
    weights["terrain"] * terrain_norm
)

# Step 6: Save the SAW result as a GeoTIFF
transform = from_origin(x_min, y_max, cell_size, cell_size)
with rasterio.open(
    "saw_probability_map.tif",
    "w",
    driver="GTiff",
    height=saw_result.shape[0],
    width=saw_result.shape[1],
    count=1,
    dtype=saw_result.dtype,
    crs="EPSG:4326",
    transform=transform,
) as dst:
    dst.write(saw_result, 1)

print("SAW probability map saved as 'saw_probability_map.tif'")

```

SAW probability map saved as 'saw\_probability\_map.tif'

```
In [26]: with rasterio.open("synthetic_lkp_raster.tif") as lkp_src:
         lkp_layer = lkp_src.read(1)
```

```
In [29]: # Grid properties
width, height = 100, 100 # Grid size
cell_size = 0.01 # Cell size in degrees
x_min, y_max = 29.0, 39.8 # Top-left corner of the grid
transform = from_origin(x_min, y_max, cell_size, cell_size)

# Create synthetic raster layers
# LKP Layer (high probability near center)
x_center, y_center = 29.4, 39.5
x = np.linspace(x_min, x_min + width * cell_size, width)
y = np.linspace(y_max, y_max - height * cell_size, height)
xv, yv = np.meshgrid(x, y)
lkp_layer = np.exp(-np.sqrt((xv - x_center) ** 2 + (yv - y_center) ** 2) / 0

# Settlement Layer (higher near specific points)
settlement_layer = np.zeros((height, width))
settlement_coords = [(29.3, 39.4), (29.5, 39.3)]
for sx, sy in settlement_coords:
    settlement_layer += np.exp(-np.sqrt((xv - sx) ** 2 + (yv - sy) ** 2) / 0

# Terrain Layer (random values)
terrain_layer = np.random.rand(height, width)

# Save layers as GeoTIFF files
layers = {
    "synthetic_lkp_raster.tif": lkp_layer,
    "synthetic_settlement_raster.tif": settlement_layer,
    "synthetic_terrain_raster.tif": terrain_layer,
}

for filename, layer in layers.items():
    with rasterio.open(
        filename,
        "w",
        driver="GTiff",
        height=layer.shape[0],
        width=layer.shape[1],
        count=1,
        dtype=layer.dtype,
        crs="EPSG:4326",
        transform=transform,
    ) as dst:
        dst.write(layer, 1)

print("Synthetic raster files created:")
for filename in layers.keys():
    print(f"- {filename}")
```

Synthetic raster files created:

- synthetic\_lkp\_raster.tif
- synthetic\_settlement\_raster.tif
- synthetic\_terrain\_raster.tif



```
In [30]: with rasterio.open("synthetic_settlement_raster.tif") as settlement_src:
          settlement_layer = settlement_src.read(1)

          with rasterio.open("synthetic_terrain_raster.tif") as terrain_src:
              terrain_layer = terrain_src.read(1)
```

```
In [32]: with rasterio.open("synthetic_lkp_raster.tif") as src:
          print(src.meta)

{'driver': 'GTiff', 'dtype': 'float64', 'nodata': None, 'width': 100, 'height': 100, 'count': 1, 'crs': CRS.from_wkt('GEOGCS["WGS 84", DATUM["WGS_1984", SPHEROID["WGS 84", 6378137, 298.257223563, AUTHORITY["EPSG", "7030"]], AUTHORITY["EPSG", "6326"]], PRIMEM["Greenwich", 0, AUTHORITY["EPSG", "8901"]], UNIT["degree", 0.0174532925199433, AUTHORITY["EPSG", "9122"]], AXIS["Latitude", NORTH], AXIS["Longitude", EAST], AUTHORITY["EPSG", "4326"]]), 'transform': Affine(0.01, 0.0, 29.0, 0.0, -0.01, 39.8)}
```

```
In [34]: def reproject_raster(input_path, output_path, target_crs):
          with rasterio.open(input_path) as src:
              transform, width, height = calculate_default_transform(
                  src.crs, target_crs, src.width, src.height, *src.bounds
              )
              kwargs = src.meta.copy()
              kwargs.update({
                  'crs': target_crs,
                  'transform': transform,
                  'width': width,
                  'height': height
              })

          with rasterio.open(output_path, 'w', **kwargs) as dst:
              for i in range(1, src.count + 1):
                  reproject(
                      source=rasterio.band(src, i),
                      destination=rasterio.band(dst, i),
                      src_transform=src.transform,
                      src_crs=src.crs,
                      dst_transform=transform,
                      dst_crs=target_crs,
                      resampling=Resampling.nearest
                  )
          reproject_raster("synthetic_lkp_raster.tif", "reprojected_lkp.tif", "EPSG:4326")
```

```
In [35]: def normalize_raster(layer):
          return (layer - np.nanmin(layer)) / (np.nanmax(layer) - np.nanmin(layer))
```

```
In [36]: weights = {
    "lkp": 0.5,          # LKP influence
    "settlement": 0.3,   # Settlement proximity
    "terrain": 0.2       # Terrain accessibility
}

saw_result = (
    weights["lkp"] * lkp_norm +
    weights["settlement"] * settlement_norm +
    weights["terrain"] * terrain_norm
)
```

```
In [38]: transform = from_origin(x_min, y_max, cell_size, cell_size)
with rasterio.open(
    "saw_probability_map.tif",
    "w",
    driver="GTiff",
    height=saw_result.shape[0],
    width=saw_result.shape[1],
    count=1,
    dtype=saw_result.dtype,
    crs="EPSG:4326",
    transform=transform,
) as dst:
    dst.write(saw_result, 1)

print("SAW probability map saved as 'saw_probability_map.tif'")

SAW probability map saved as 'saw_probability_map.tif'
```

```

In [39]: # Load rasters
with rasterio.open("synthetic_lkp_raster.tif") as lkp_src:
    lkp_layer = lkp_src.read(1)

with rasterio.open("synthetic_settlement_raster.tif") as settlement_src:
    settlement_layer = settlement_src.read(1)

with rasterio.open("synthetic_terrain_raster.tif") as terrain_src:
    terrain_layer = terrain_src.read(1)

# Normalize layers
lkp_norm = normalize_raster(lkp_layer)
settlement_norm = normalize_raster(settlement_layer)
terrain_norm = normalize_raster(terrain_layer)

# Combine layers using SAW
weights = {"lkp": 0.5, "settlement": 0.3, "terrain": 0.2}
saw_result = (
    weights["lkp"] * lkp_norm +
    weights["settlement"] * settlement_norm +
    weights["terrain"] * terrain_norm
)

# Save the combined map
with rasterio.open(
    "saw_probability_map.tif",
    "w",
    driver="GTiff",
    height=saw_result.shape[0],
    width=saw_result.shape[1],
    count=1,
    dtype=saw_result.dtype,
    crs=lkp_src.crs,
    transform=lkp_src.transform,
) as dst:
    dst.write(saw_result, 1)

print("SAW probability map saved as 'saw_probability_map.tif'")

```

SAW probability map saved as 'saw\_probability\_map.tif'

```

In [40]: # Step 1: Load Existing Layers (LKP, Settlements, Terrain)
with rasterio.open("synthetic_lkp_raster.tif") as lkp_src:
    lkp_layer = lkp_src.read(1)
    transform = lkp_src.transform
    crs = lkp_src.crs

with rasterio.open("synthetic_settlement_raster.tif") as settlement_src:
    settlement_layer = settlement_src.read(1)

with rasterio.open("synthetic_terrain_raster.tif") as terrain_src:
    terrain_layer = terrain_src.read(1)

# Step 2: Generate Synthetic Layers for New Criteria
# Accessibility Layer (proximity to roads)
accessibility_layer = np.zeros_like(lkp_layer)

```

```

road_coords = [(29.3, 39.4), (29.5, 39.3)]
for rx, ry in road_coords:
    xv, yv = np.meshgrid(
        np.linspace(29.0, 30.0, accessibility_layer.shape[1]),
        np.linspace(39.0, 40.0, accessibility_layer.shape[0])
    )
    accessibility_layer += np.exp(-np.sqrt((xv - rx)**2 + (yv - ry)**2) / 0.

# Weather Layer (random suitability values)
weather_layer = np.random.rand(*lkp_layer.shape)

# Infrastructure Layer (distance to airports)
infrastructure_layer = np.zeros_like(lkp_layer)
airport_coords = [(29.2, 39.2), (29.6, 39.6)]
for ax, ay in airport_coords:
    infrastructure_layer += np.exp(-np.sqrt((xv - ax)**2 + (yv - ay)**2) / 0.

# Step 3: Normalize All Layers
def normalize(layer):
    return (layer - np.nanmin(layer)) / (np.nanmax(layer) - np.nanmin(layer))

lkp_norm = normalize(lkp_layer)
settlement_norm = normalize(settlement_layer)
terrain_norm = normalize(terrain_layer)
accessibility_norm = normalize(accessibility_layer)
weather_norm = normalize(weather_layer)
infrastructure_norm = normalize(infrastructure_layer)

# Step 4: Define Weights for All Criteria
weights = {
    "lkp": 0.4,
    "settlement": 0.2,
    "terrain": 0.1,
    "accessibility": 0.1,
    "weather": 0.1,
    "infrastructure": 0.1,
}

# Step 5: Compute Weighted Overlay (SAW)
saw_result = (
    weights["lkp"] * lkp_norm +
    weights["settlement"] * settlement_norm +
    weights["terrain"] * terrain_norm +
    weights["accessibility"] * accessibility_norm +
    weights["weather"] * weather_norm +
    weights["infrastructure"] * infrastructure_norm
)

# Step 6: Save the SAW Probability Map
output_file = "enhanced_saw_probability_map.tif"
with rasterio.open(
    output_file,
    "w",
    driver="GTiff",
    height=saw_result.shape[0],
    width=saw_result.shape[1],
    count=1,

```

```

dtype=saw_result.dtype,
crs=crs,
transform=transform,
) as dst:
    dst.write(saw_result, 1)

print(f"Enhanced SAW probability map saved as '{output_file}'")

```

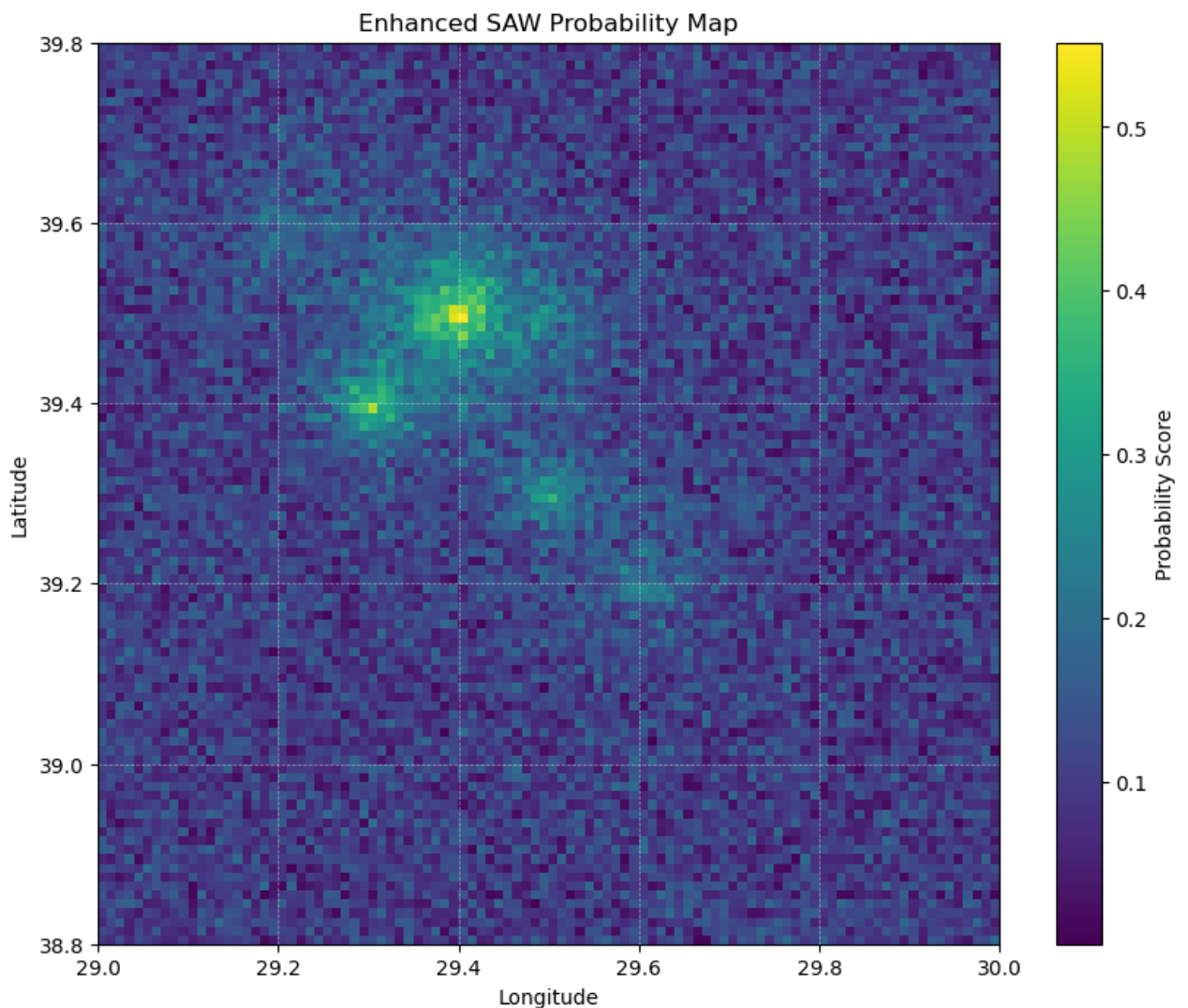
Enhanced SAW probability map saved as 'enhanced\_saw\_probability\_map.tif'

```

In [42]: # Load the SAW probability map
with rasterio.open("enhanced_saw_probability_map.tif") as src:
    saw_map = src.read(1)
    extent = [src.bounds.left, src.bounds.right, src.bounds.bottom, src.boun

# Plot the probability map
plt.figure(figsize=(10, 8))
plt.imshow(saw_map, extent=extent, cmap="viridis", origin="upper")
plt.colorbar(label="Probability Score")
plt.title("Enhanced SAW Probability Map")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.grid(color='white', linestyle='--', linewidth=0.5, alpha=0.5)
plt.show()

```



```

In [47]: # Load the SAW probability map
with rasterio.open("enhanced_saw_probability_map.tif") as src:
    saw_map = src.read(1) # Read the first band
    bounds = [[src.bounds.bottom, src.bounds.left], [src.bounds.top, src.bou

# Normalize data for Folium
saw_map_normalized = (saw_map - np.nanmin(saw_map)) / (np.nanmax(saw_map) -

# Convert the normalized array to a valid Folium input (scale to [0, 255])
saw_map_folium = (saw_map_normalized * 255).astype(np.uint8)

# Create a Folium map
m = folium.Map(
    location=[
        (bounds[0][0] + bounds[1][0]) / 2,
        (bounds[0][1] + bounds[1][1]) / 2
    ],
    zoom_start=10
)

# Overlay the probability map
folium.raster_layers.ImageOverlay(
    image=saw_map_folium,
    bounds=bounds,
    colormap=lambda x: (1, 0, 0, x), # Red with transparency
    name="SAW Probability Map"
).add_to(m)

# Add layer control and save the map
folium.LayerControl().add_to(m)
m.save("saw_probability_map.html")
print("Map saved as 'saw_probability_map.html'")

```

Map saved as 'saw\_probability\_map.html'

```

In [51]: # Step 1: Define Criteria and Pairwise Comparison Matrix
criteria = ["LKP", "Settlement", "Terrain", "Accessibility", "Weather"]
pairwise_matrix = np.array([
    [1, 3, 5, 7, 9], # LKP compared to others
    [1/3, 1, 3, 5, 7], # Settlement compared to others
    [1/5, 1/3, 1, 3, 5], # Terrain compared to others
    [1/7, 1/5, 1/3, 1, 3], # Accessibility compared to others
    [1/9, 1/7, 1/5, 1/3, 1] # Weather compared to others
])

# Step 2: Normalize the Pairwise Matrix
column_sums = pairwise_matrix.sum(axis=0)
normalized_matrix = pairwise_matrix / column_sums

# Step 3: Calculate Criteria Weights
criteria_weights = normalized_matrix.mean(axis=1)

# Step 4: Display the Results
weights_df = pd.DataFrame({
    "Criteria": criteria,
    "Weight": criteria_weights
}).sort_values(by="Weight", ascending=False)

print("Criteria Weights:")
print(weights_df)

# Step 5: Check Consistency
# Calculate Consistency Index (CI) and Consistency Ratio (CR)
eigenvalues = np.dot(pairwise_matrix, criteria_weights) / criteria_weights
lambda_max = eigenvalues.mean()
n = pairwise_matrix.shape[0]
ci = (lambda_max - n) / (n - 1)
random_index = {1: 0.00, 2: 0.00, 3: 0.58, 4: 0.90, 5: 1.12, 6: 1.24, 7: 1.32}
cr = ci / random_index[n]

print(f"\nConsistency Index (CI): {ci:.4f}")
print(f"Consistency Ratio (CR): {cr:.4f}")

if cr < 0.1:
    print("The pairwise comparisons are consistent.")
else:
    print("The pairwise comparisons are not consistent. Reassess the matrix.")

```

Criteria Weights:

	Criteria	Weight
0	LKP	0.502819
1	Settlement	0.260232
2	Terrain	0.134350
3	Accessibility	0.067778
4	Weather	0.034821

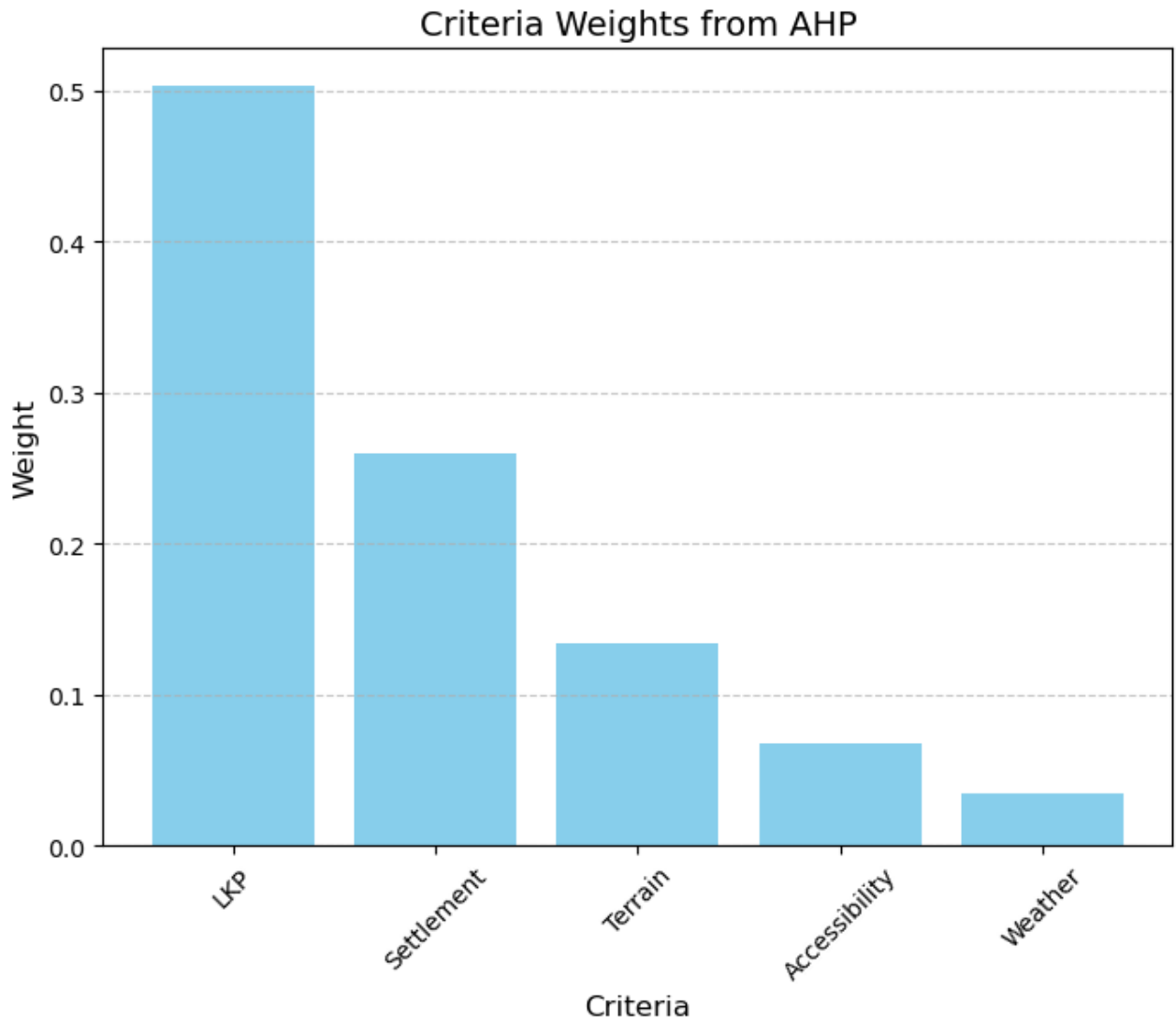
Consistency Index (CI): 0.0607

Consistency Ratio (CR): 0.0542

The pairwise comparisons are consistent.

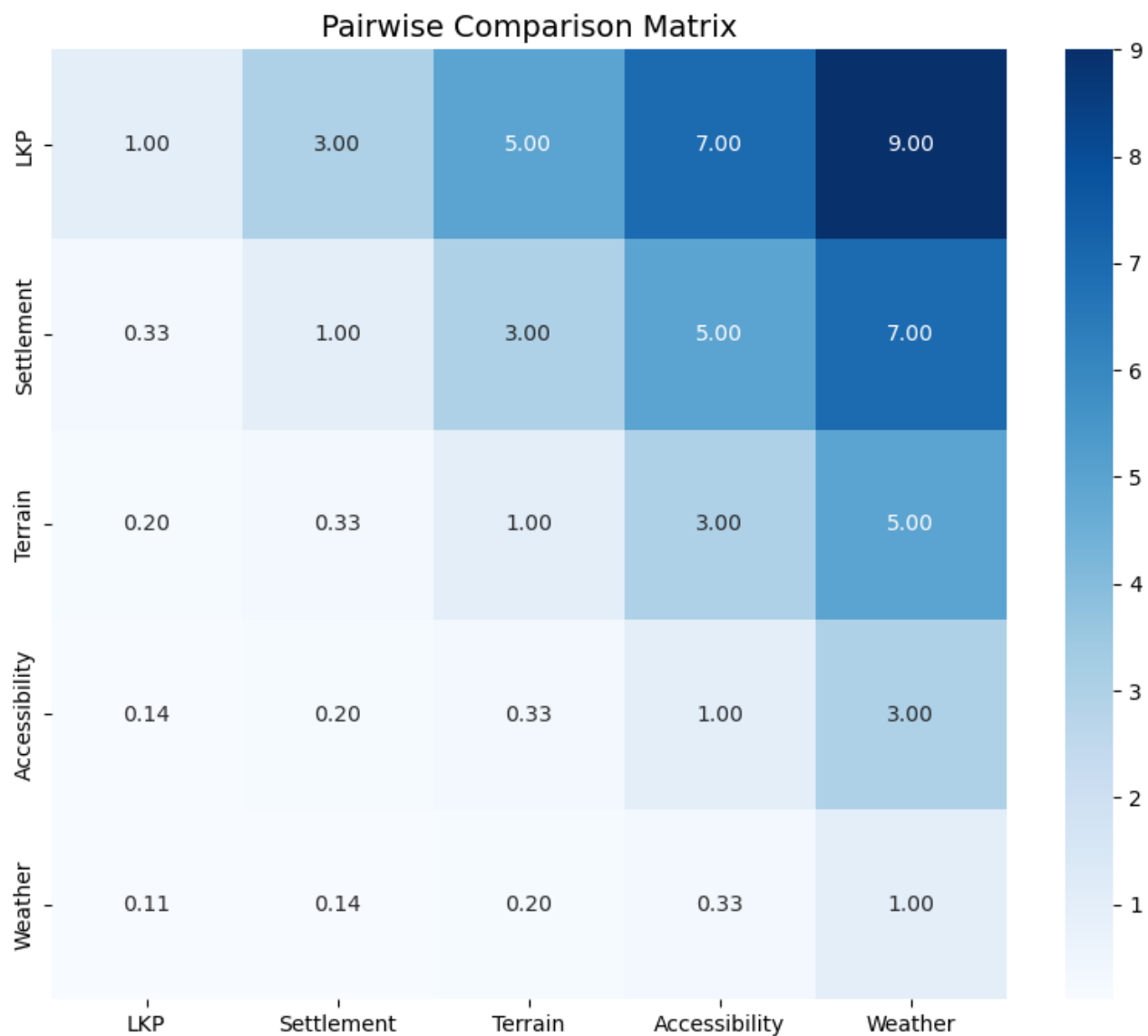


```
In [52]: # Visualization: Criteria Weights
plt.figure(figsize=(8, 6))
plt.bar(weights_df["Criteria"], weights_df["Weight"], color="skyblue")
plt.title("Criteria Weights from AHP", fontsize=14)
plt.xlabel("Criteria", fontsize=12)
plt.ylabel("Weight", fontsize=12)
plt.xticks(rotation=45)
plt.grid(axis="y", linestyle="--", alpha=0.7)
plt.show()
```



```
In [54]: # Convert Pairwise Matrix to DataFrame for Visualization
pairwise_df = pd.DataFrame(pairwise_matrix, index=criteria, columns=criteria)

# Visualization: Heatmap of Pairwise Comparisons
plt.figure(figsize=(10, 8))
sns.heatmap(pairwise_df, annot=True, fmt=".2f", cmap="Blues", cbar=True)
plt.title("Pairwise Comparison Matrix", fontsize=14)
plt.show()
```



```
In [56]: # Function to Draw a Gauge
def plot_gauge(cr, threshold=0.1):
    fig, ax = plt.subplots(figsize=(6, 4), subplot_kw={"aspect": "equal"})

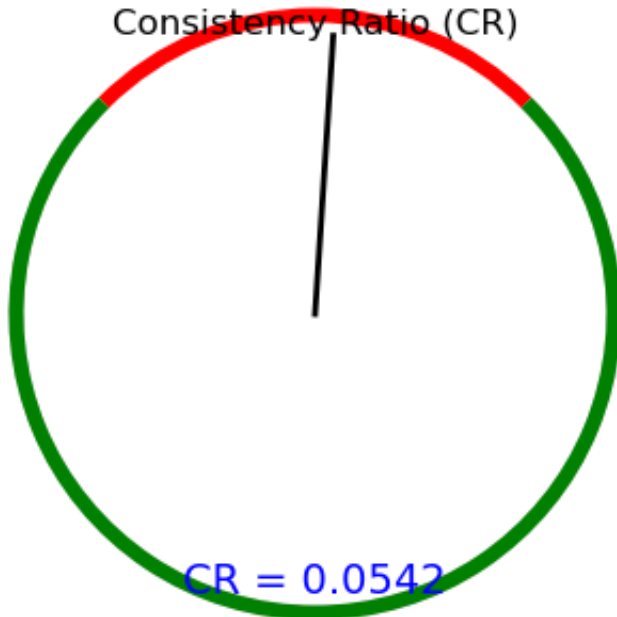
    # Draw the base arc
    theta1, theta2 = 135, 45
    ax.add_patch(patches.Arc((0.5, 0.5), 1.5, 1.5, theta1=theta1, theta2=theta2))

    # Threshold marker
    angle_threshold = 135 - (threshold / 0.1 * 90)
    ax.add_patch(patches.Arc((0.5, 0.5), 1.5, 1.5, theta1=angle_threshold, theta2=angle_threshold))
    ax.add_patch(patches.Arc((0.5, 0.5), 1.5, 1.5, theta1=135, theta2=angle_threshold))

    # CR marker
    angle_cr = 135 - (cr / 0.1 * 90)
    ax.plot([0.5, 0.5 + 0.7 * np.cos(np.radians(angle_cr))],
            [0.5, 0.5 + 0.7 * np.sin(np.radians(angle_cr))],
            color="black", lw=2)

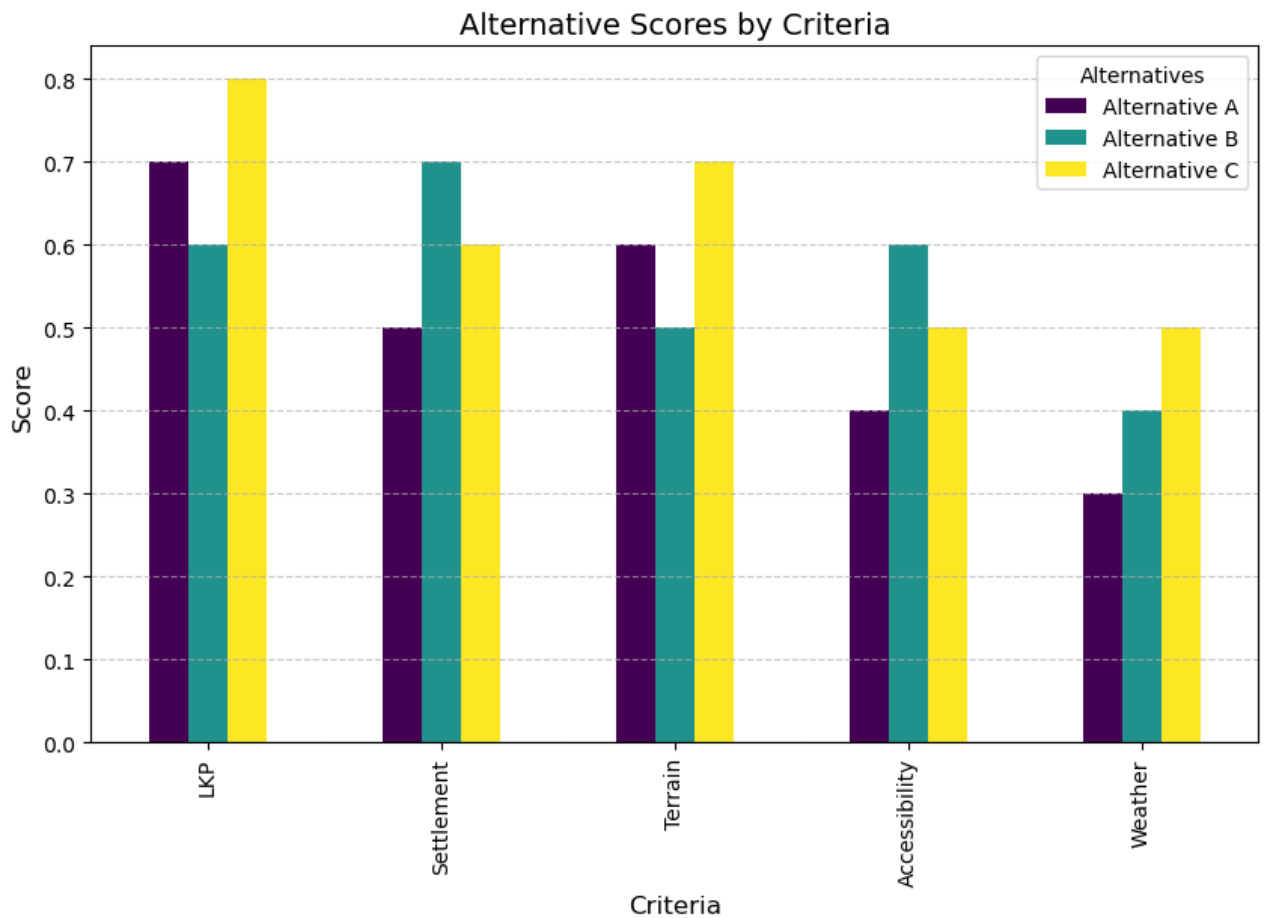
    # Annotations
    ax.text(0.5, 1.2, "Consistency Ratio (CR)", ha="center", fontsize=12)
    ax.text(0.5, -0.2, f"CR = {cr:.4f}", ha="center", fontsize=14, color="blue")
    ax.axis("off")
    plt.show()

# Visualize CR Gauge
plot_gauge(cr)
```



```
In [57]: # Example Data (Weights for 3 Alternatives)
alternative_scores = pd.DataFrame({
    "Criteria": ["LKP", "Settlement", "Terrain", "Accessibility", "Weather"],
    "Alternative A": [0.7, 0.5, 0.6, 0.4, 0.3],
    "Alternative B": [0.6, 0.7, 0.5, 0.6, 0.4],
    "Alternative C": [0.8, 0.6, 0.7, 0.5, 0.5],
}).set_index("Criteria")

# Plot
alternative_scores.plot(kind="bar", figsize=(10, 6), colormap="viridis")
plt.title("Alternative Scores by Criteria", fontsize=14)
plt.ylabel("Score", fontsize=12)
plt.xlabel("Criteria", fontsize=12)
plt.legend(title="Alternatives")
plt.grid(axis="y", linestyle="--", alpha=0.7)
plt.show()
```



```

In [58]: # Step 1: Define Criteria Values for an Example Alternative
# Example: Normalized values for one grid cell or region
criteria = ["LKP", "Settlement", "Terrain", "Accessibility", "Weather"]
values = np.array([0.9, 0.7, 0.5, 0.3, 0.2]) # Normalized values (example)

# Step 2: Sort Criteria Values in Descending Order
sorted_values = np.sort(values)[::-1] # Descending order

# Step 3: Define OWA Weights
# Example: Neutral aggregation (equal weights)
owa_weights = np.array([0.2, 0.2, 0.2, 0.2, 0.2]) # Must sum to 1

# Adjust OWA weights for optimism or pessimism
# Optimistic: [1, 0, 0, 0, 0] (focus on the highest value)
# Pessimistic: [0, 0, 0, 0, 1] (focus on the lowest value)

# Step 4: Compute OWA Score
owa_score = np.sum(owa_weights * sorted_values)

# Display Results
print(f"Criteria: {criteria}")
print(f"Original Values: {values}")
print(f"Sorted Values: {sorted_values}")
print(f"OWA Weights: {owa_weights}")
print(f"OWA Score: {owa_score:.4f}")

Criteria: ['LKP', 'Settlement', 'Terrain', 'Accessibility', 'Weather']
Original Values: [0.9 0.7 0.5 0.3 0.2]
Sorted Values: [0.9 0.7 0.5 0.3 0.2]
OWA Weights: [0.2 0.2 0.2 0.2 0.2]
OWA Score: 0.5200

```

```

In [59]: # Load Criteria Layers (Rasters)
with rasterio.open("synthetic_lkp_raster.tif") as lkp_src:
    lkp_layer = lkp_src.read(1) # Read the first band

with rasterio.open("synthetic_settlement_raster.tif") as settlement_src:
    settlement_layer = settlement_src.read(1)

with rasterio.open("synthetic_terrain_raster.tif") as terrain_src:
    terrain_layer = terrain_src.read(1)

# Combine Layers into a 3D Array (Criteria Layers)
criteria_layers = np.stack([lkp_layer, settlement_layer, terrain_layer], axis=0)

# Normalize Layers to [0, 1]
criteria_layers = (criteria_layers - criteria_layers.min(axis=(1, 2), keepdims=True)
                  * criteria_layers.max(axis=(1, 2), keepdims=True) - criteria_layers.min(axis=(1, 2), keepdims=True))

# Define OWA Weights (Neutral Example)
owa_weights = np.array([0.5, 0.3, 0.2]) # Adjust based on strategy (must sum to 1)

# Compute OWA Score for Each Cell
owa_scores = np.zeros((criteria_layers.shape[1], criteria_layers.shape[2]))
for i in range(criteria_layers.shape[1]):
    for j in range(criteria_layers.shape[2]):
        sorted_values = np.sort(criteria_layers[:, i, j])[::-1]
        owa_scores[i, j] = np.sum(owa_weights * sorted_values)

# Save OWA Scores as GeoTIFF
with rasterio.open(
    "owa_probability_map.tif",
    "w",
    driver="GTiff",
    height=owa_scores.shape[0],
    width=owa_scores.shape[1],
    count=1,
    dtype=owa_scores.dtype,
    crs=lkp_src.crs,
    transform=lkp_src.transform,
) as dst:
    dst.write(owa_scores, 1)

print("OWA probability map saved as 'owa_probability_map.tif'")

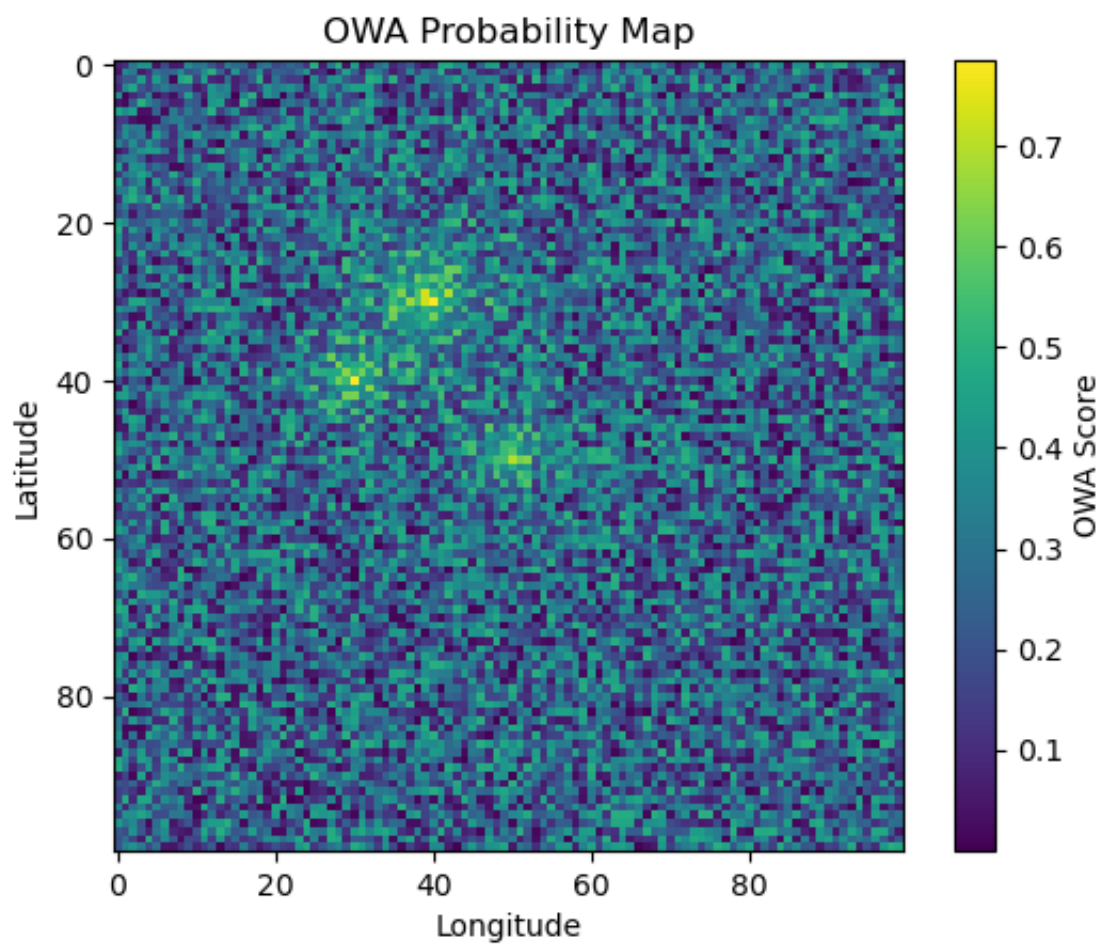
OWA probability map saved as 'owa_probability_map.tif'

```

```

In [60]: plt.imshow(owa_scores, cmap="viridis", origin="upper")
plt.colorbar(label="OWA Score")
plt.title("OWA Probability Map")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.show()

```





```

In [61]: # Load the OWA Probability Map
with rasterio.open("owa_probability_map.tif") as src:
    owa_map = src.read(1) # Read the first band
    bounds = [[src.bounds.bottom, src.bounds.left], [src.bounds.top, src.bou

# Normalize the OWA map for display in Folium
owa_map_normalized = (owa_map - np.nanmin(owa_map)) / (np.nanmax(owa_map) -
owa_map_normalized = (owa_map_normalized * 255).astype(np.uint8) # Scale to

# Create a Folium Map
m = folium.Map(
    location=[(bounds[0][0] + bounds[1][0]) / 2, (bounds[0][1] + bounds[1][1]) / 2],
    zoom_start=10,
    tiles="cartodbpositron"
)

# Overlay the OWA map
folium.raster_layers.ImageOverlay(
    image=owa_map_normalized,
    bounds=bounds,
    colormap=lambda x: (x, 0, 1 - x, 0.6), # Blue gradient with transparency
    name="OWA Probability Map"
).add_to(m)

# Add a layer control and save the map
folium.LayerControl().add_to(m)
m.save("owa_probability_map.html")
print("Map saved as 'owa_probability_map.html'")

```

Map saved as 'owa\_probability\_map.html'

```
In [63]: # Step 1: Load the Probability Map
with rasterio.open("owa_probability_map.tif") as src:
    probability_map = src.read(1) # Read the first band
    transform = src.transform
    crs = src.crs

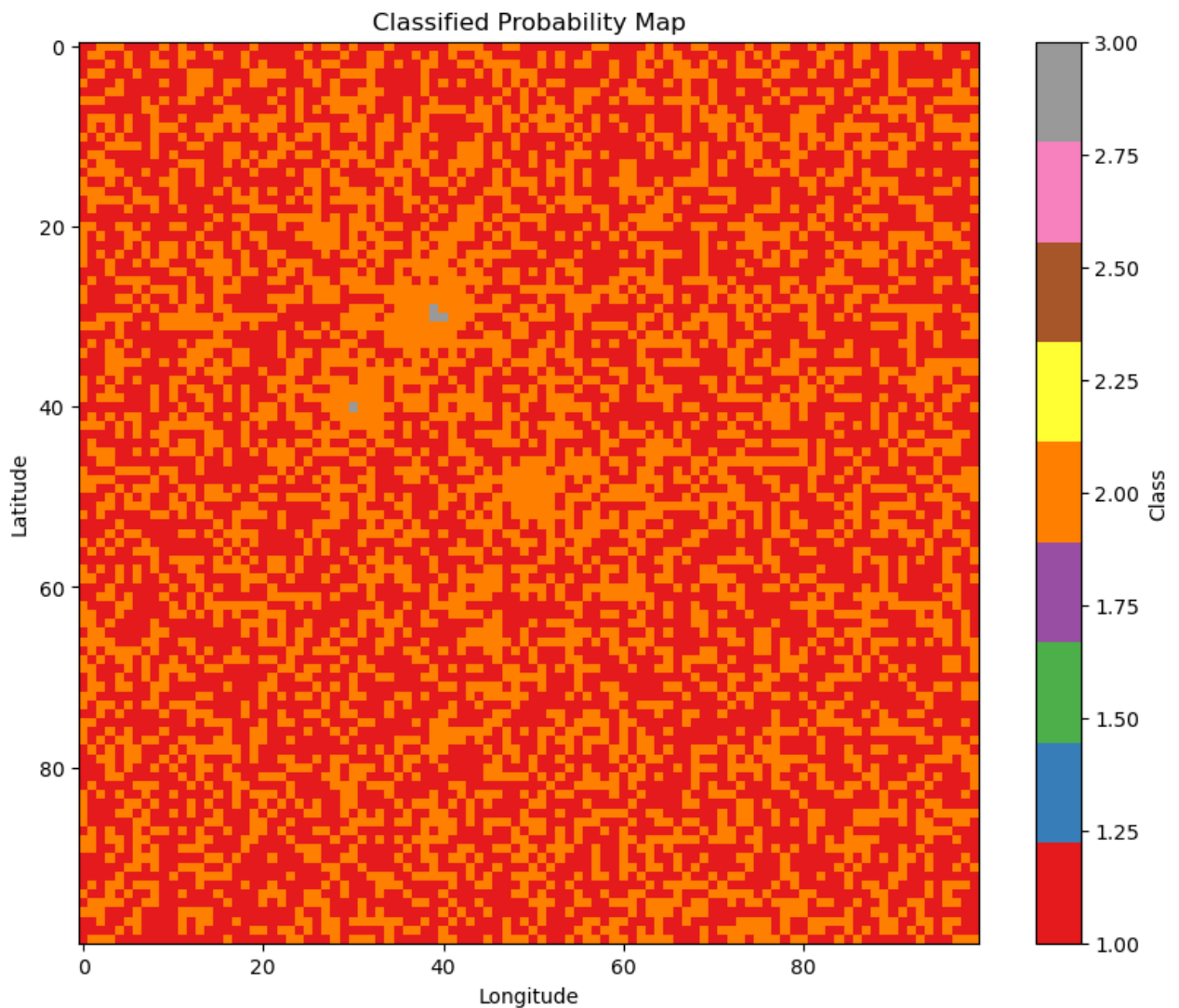
# Step 2: Define Classification Thresholds
# Example: Custom thresholds for 3 classes (Low, Medium, High)
thresholds = [0.0, 0.3, 0.7, 1.0] # Adjust as needed

# Step 3: Classify the Map
classified_map = np.digitize(probability_map, bins=thresholds, right=True)

# Step 4: Visualize the Classified Map
plt.figure(figsize=(10, 8))
plt.imshow(classified_map, cmap="Set1", origin="upper")
plt.colorbar(label="Class")
plt.title("Classified Probability Map")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.show()

# Step 5: Save the Classified Map
output_file = "classified_probability_map.tif"
with rasterio.open(
    output_file,
    "w",
    driver="GTiff",
    height=classified_map.shape[0],
    width=classified_map.shape[1],
    count=1,
    dtype=classified_map.dtype,
    crs=crs,
    transform=transform,
) as dst:
    dst.write(classified_map, 1)

print(f"Classified probability map saved as '{output_file}')
```



Classified probability map saved as 'classified\_probability\_map.tif'

```
In [64]: thresholds = np.linspace(np.min(probability_map), np.max(probability_map), n
```

```
In [65]: thresholds = np.percentile(probability_map[~np.isnan(probability_map)], [0,
```

```
In [71]: # Flatten the probability map, excluding NaN values
data = probability_map[~np.isnan(probability_map)].flatten()

# Specify the number of classes
num_classes = 3

# Calculate Jenks natural breaks
thresholds = jenkspy.jenks_breaks(data, num_classes)

print("Jenks Natural Breaks Thresholds:", thresholds)

Jenks Natural Breaks Thresholds: [0.00014358225047216507, 0.1766282378287802
6, 0.3456137984942614, 0.7852180498304305]
```

```
In [72]: classified_map = np.digitize(probability_map, bins=thresholds, right=True)
```

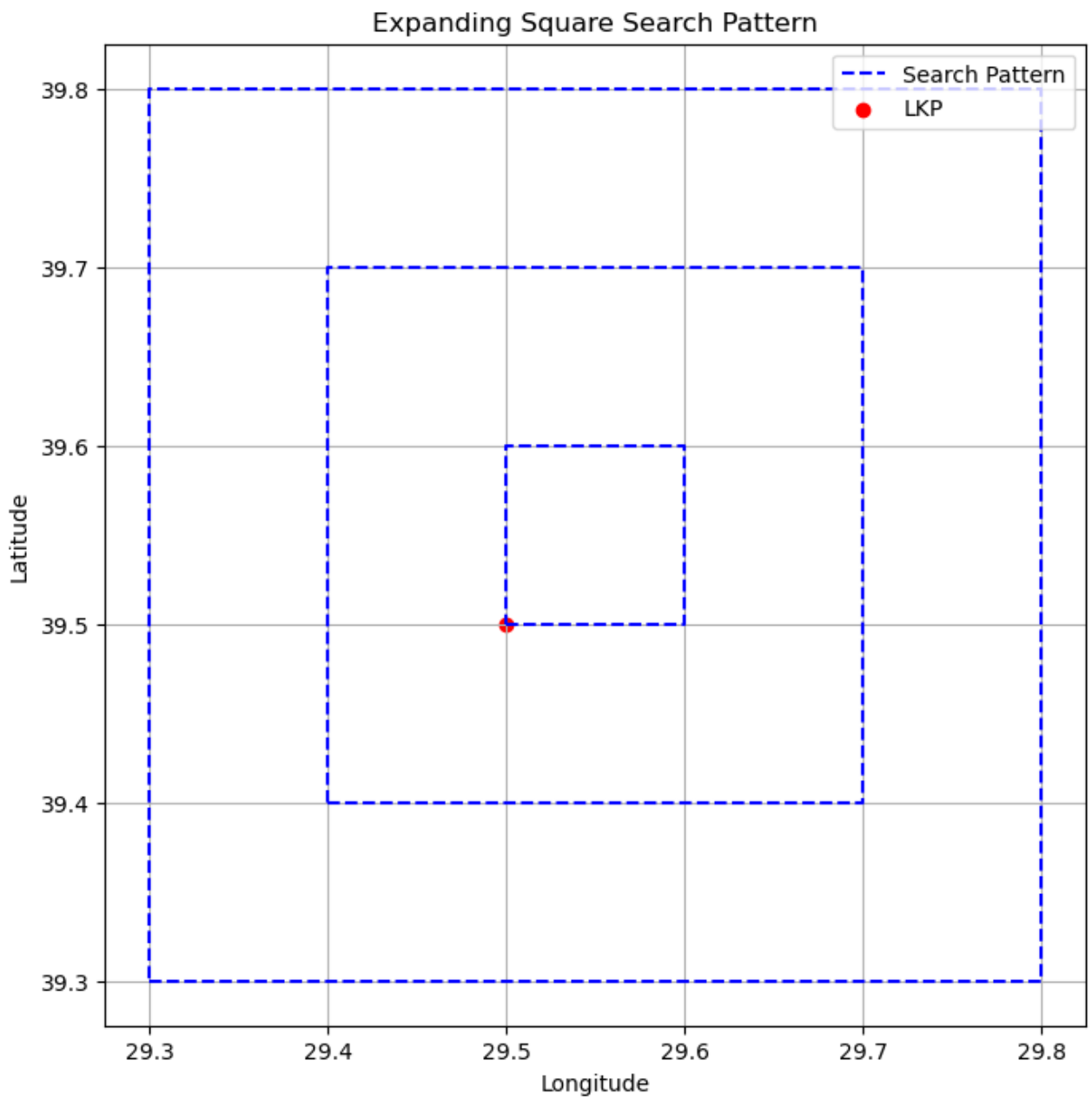
```
In [73]: import matplotlib.pyplot as plt
from shapely.geometry import Point, LineString, Polygon
import geopandas as gpd

# Define the Last Known Position (LKP)
lkp = Point(29.5, 39.5)

# Create Expanding Square Search Pattern
square_coords = [
    [(29.5, 39.5), (29.6, 39.5), (29.6, 39.6), (29.5, 39.6), (29.5, 39.5)],
    [(29.4, 39.4), (29.7, 39.4), (29.7, 39.7), (29.4, 39.7), (29.4, 39.4)],
    [(29.3, 39.3), (29.8, 39.3), (29.8, 39.8), (29.3, 39.8), (29.3, 39.3)],
]

# Convert to GeoDataFrame
squares = [Polygon(coords) for coords in square_coords]
search_pattern_gdf = gpd.GeoDataFrame(geometry=squares, crs="EPSG:4326")

# Plot the Search Pattern
fig, ax = plt.subplots(figsize=(10, 8))
search_pattern_gdf.boundary.plot(ax=ax, color="blue", linestyle="--", label="Search Pattern")
gpd.GeoSeries([lkp]).plot(ax=ax, color="red", label="LKP")
plt.title("Expanding Square Search Pattern")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.legend()
plt.grid()
plt.show()
```



```
In [74]: # Step 1: Load Criteria Layers
with rasterio.open("synthetic_lkp_raster.tif") as lkp_src:
    lkp_layer = lkp_src.read(1)

with rasterio.open("synthetic_settlement_raster.tif") as settlement_src:
    settlement_layer = settlement_src.read(1)

with rasterio.open("synthetic_terrain_raster.tif") as terrain_src:
    terrain_layer = terrain_src.read(1)

# Step 2: Normalize Criteria Layers (0 to 1 scale)
def normalize(layer):
    return (layer - np.nanmin(layer)) / (np.nanmax(layer) - np.nanmin(layer))

lkp_norm = normalize(lkp_layer)
settlement_norm = normalize(settlement_layer)
```

```

terrain_norm = normalize(terrain_layer)

# Step 3: Define Weights for Each Criterion
weights = {
    "lkp": 0.5,          # LKP Proximity
    "settlement": 0.3,   # Settlement Proximity
    "terrain": 0.2       # Terrain Accessibility
}

# Step 4: Compute Weighted Sum
probability_map = (
    weights["lkp"] * lkp_norm +
    weights["settlement"] * settlement_norm +
    weights["terrain"] * terrain_norm
)

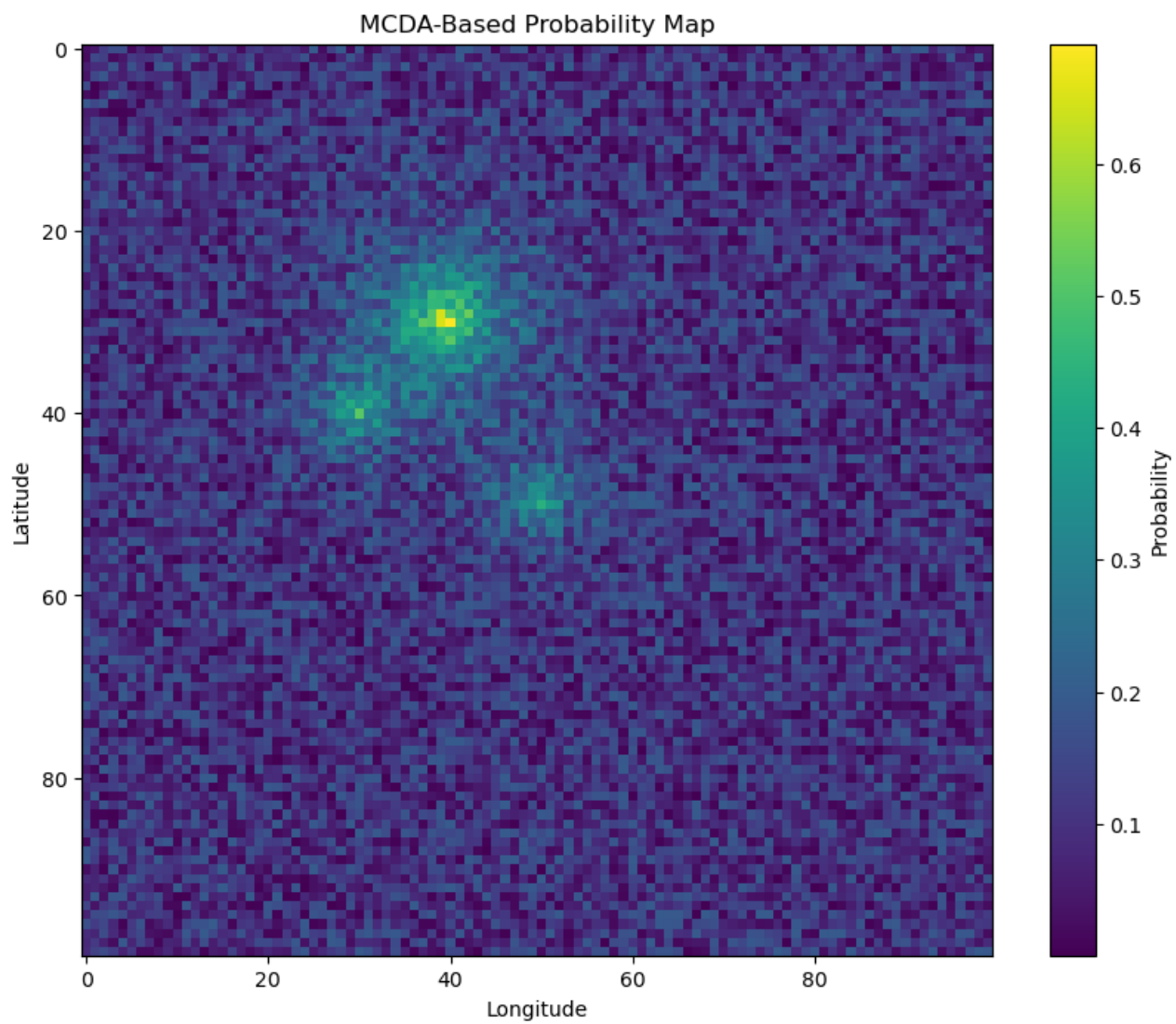
# Step 5: Save Probability Map as GeoTIFF
transform = lkp_src.transform
crs = lkp_src.crs
with rasterio.open(
    "mcda_probability_map.tif",
    "w",
    driver="GTiff",
    height=probability_map.shape[0],
    width=probability_map.shape[1],
    count=1,
    dtype=probability_map.dtype,
    crs=crs,
    transform=transform,
) as dst:
    dst.write(probability_map, 1)

print("MCDA-based probability map saved as 'mcda_probability_map.tif'")

# Step 6: Visualize Probability Map
plt.figure(figsize=(10, 8))
plt.imshow(probability_map, cmap="viridis", origin="upper")
plt.colorbar(label="Probability")
plt.title("MCDA-Based Probability Map")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.show()

```

MCDA-based probability map saved as 'mcda\_probability\_map.tif'





```

In [75]: # Step 1: Load Probability Map
with rasterio.open("mcda_probability_map.tif") as src:
    probability_map = src.read(1)
    transform = src.transform
    crs = src.crs

# Step 2: Define Thresholds
# Define thresholds for Low, Medium, High
thresholds = [0.0, 0.3, 0.7, 1.0] # Low: 0-0.3, Medium: 0.3-0.7, High: 0.7-
labels = [1, 2, 3] # 1: Low, 2: Medium, 3: High

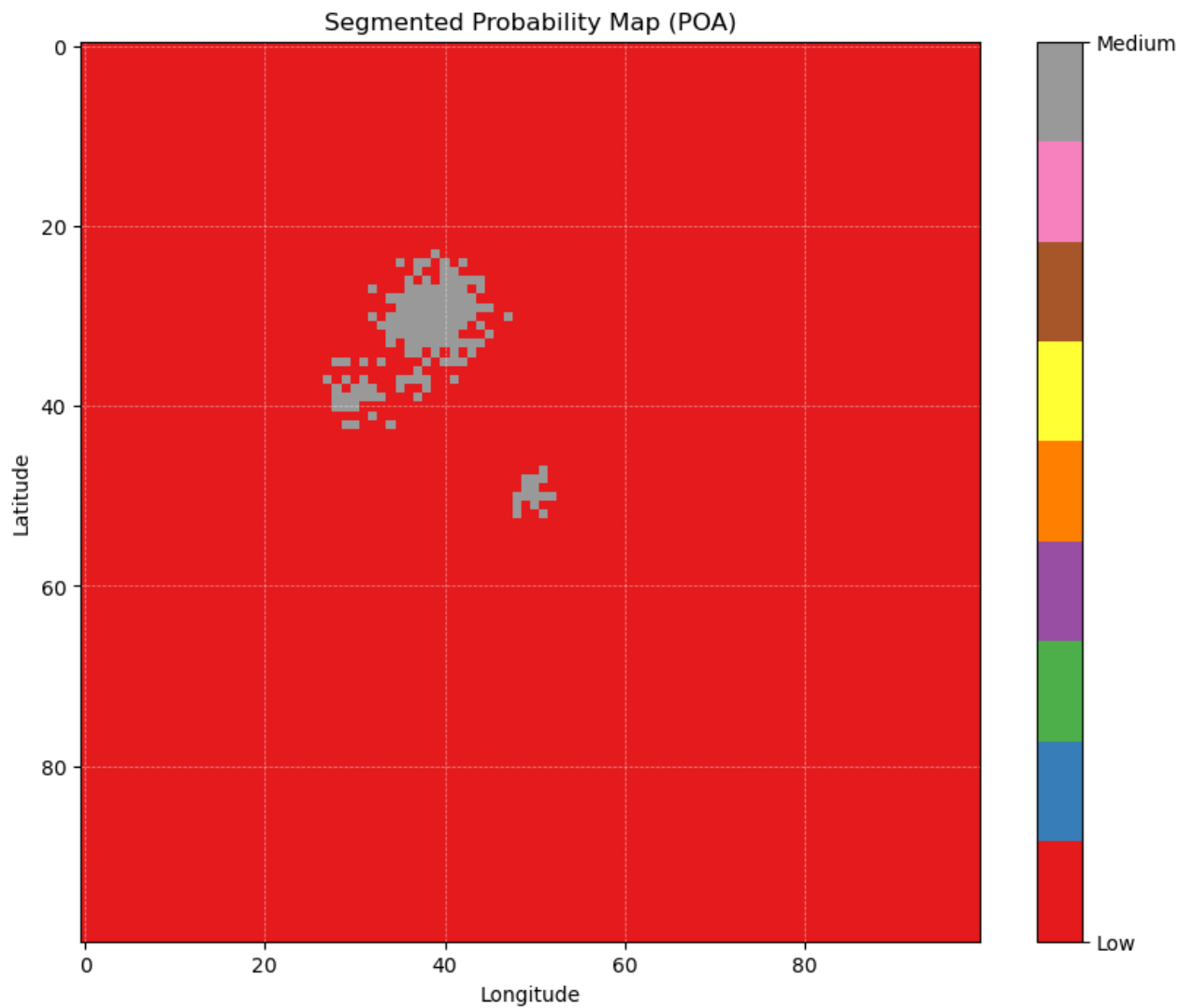
# Step 3: Segment Probability Map
segmented_map = np.digitize(probability_map, bins=thresholds, right=True)

# Step 4: Visualize Segmented Map
plt.figure(figsize=(10, 8))
plt.imshow(segmented_map, cmap="Set1", origin="upper")
cbar = plt.colorbar(ticks=[1, 2, 3])
cbar.ax.set_yticklabels(["Low", "Medium", "High"])
plt.title("Segmented Probability Map (POA)")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.grid(color="white", linestyle="--", linewidth=0.5, alpha=0.5)
plt.show()

# Step 5: Save Segmented Map as GeoTIFF
output_file = "segmented_probability_map.tif"
with rasterio.open(
    output_file,
    "w",
    driver="GTiff",
    height=segmented_map.shape[0],
    width=segmented_map.shape[1],
    count=1,
    dtype=segmented_map.dtype,
    crs=crs,
    transform=transform,
) as dst:
    dst.write(segmented_map, 1)

print(f"Segmented probability map saved as '{output_file}'")

```



Segmented probability map saved as 'segmented\_probability\_map.tif'

```
In [76]: # Step 1: Count the number of cells in each segment
unique, counts = np.unique(segmented_map, return_counts=True)
cell_counts = dict(zip(unique, counts))

# Step 2: Calculate area (assuming square grid cells)
cell_area = (transform[0] ** 2) # Cell size in map units (e.g., degrees or
area_by_class = {label: count * cell_area for label, count in cell_counts.items()}

# Display results
for label, area in area_by_class.items():
    class_name = ["Low", "Medium", "High"][label - 1]
    print(f"Class '{class_name}' covers an area of {area:.2f} square units.")
```

Class 'Low' covers an area of 0.99 square units.

Class 'Medium' covers an area of 0.01 square units.

```

In [78]: # Define Search Area and LKP
lkp = Point(29.5, 39.5)
search_area = Polygon([(29.3, 39.3), (29.7, 39.3), (29.7, 39.7), (29.3, 39.7)

# Generate Search Patterns
# 1. Expanding Square
expanding_square_coords = [
    [(29.5, 39.5), (29.6, 39.5), (29.6, 39.6), (29.5, 39.6), (29.5, 39.5)],
    [(29.4, 39.4), (29.7, 39.4), (29.7, 39.7), (29.4, 39.7), (29.4, 39.4)]
]
expanding_square = [Polygon(coords) for coords in expanding_square_coords]

# 2. Parallel Track
parallel_tracks = [
    LineString([(29.3, y), (29.7, y)]) for y in [39.3, 39.4, 39.5, 39.6, 39.7]
]

# Convert to GeoDataFrames
expanding_square_gdf = gpd.GeoDataFrame(geometry=expanding_square, crs="EPSG:4326")
parallel_track_gdf = gpd.GeoDataFrame(geometry=parallel_tracks, crs="EPSG:4326")

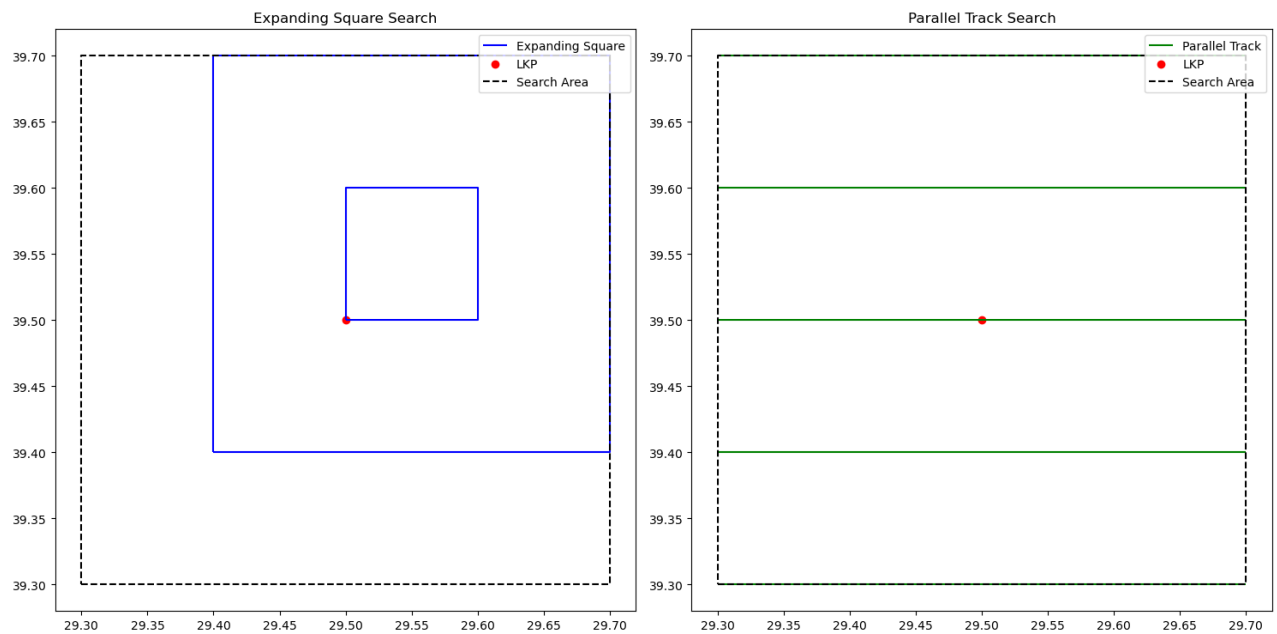
# Visualize the Search Patterns
fig, ax = plt.subplots(1, 2, figsize=(15, 8))

# Expanding Square Pattern
expanding_square_gdf.boundary.plot(ax=ax[0], color="blue", label="Expanding Square")
gpd.GeoSeries([lkp]).plot(ax=ax[0], color="red", label="LKP")
gpd.GeoSeries([search_area]).boundary.plot(ax=ax[0], color="black", linestyle="dashed")
ax[0].set_title("Expanding Square Search")
ax[0].legend()

# Parallel Track Pattern
parallel_track_gdf.plot(ax=ax[1], color="green", label="Parallel Track")
gpd.GeoSeries([lkp]).plot(ax=ax[1], color="red", label="LKP")
gpd.GeoSeries([search_area]).boundary.plot(ax=ax[1], color="black", linestyle="dashed")
ax[1].set_title("Parallel Track Search")
ax[1].legend()

plt.tight_layout()
plt.show()

```



```
In [81]: # Define coordinates for the high-probability area (example)
high_prob_coords = [
    (29.4, 39.4), (29.6, 39.4), (29.6, 39.6), (29.4, 39.6), (29.4, 39.4)
]

# Create the high-probability polygon
high_prob_area = gpd.GeoDataFrame(geometry=[Polygon(high_prob_coords)], crs=
```

```
In [82]: # Calculate the coverage of high-probability areas
expanding_square_coverage = expanding_square_gdf.intersection(high_prob_area)
parallel_track_coverage = parallel_track_gdf.intersection(high_prob_area.unary_union)

# Display the results
print(f"Expanding Square Coverage (area units): {expanding_square_coverage}")
print(f"Parallel Track Coverage (area units): {parallel_track_coverage}")
```

```
/var/folders/fp/drm5pw1d13x1n4wvd2q14v0m0000gn/T/ipykernel_26763/3804832417.
py:2: DeprecationWarning: The 'unary_union' attribute is deprecated, use the
'unary_union()' method instead.
```

```
expanding_square_coverage = expanding_square_gdf.intersection(high_prob_ar
ea.unary_union).area.sum()
```

```
/var/folders/fp/drm5pw1d13x1n4wvd2q14v0m0000gn/T/ipykernel_26763/3804832417.
py:2: UserWarning: Geometry is in a geographic CRS. Results from 'area' are
likely incorrect. Use 'GeoSeries.to_crs()' to re-project geometries to a pro
jected CRS before this operation.
```

```
expanding_square_coverage = expanding_square_gdf.intersection(high_prob_ar
ea.unary_union).area.sum()
```

```
/var/folders/fp/drm5pw1d13x1n4wvd2q14v0m0000gn/T/ipykernel_26763/3804832417.
py:3: DeprecationWarning: The 'unary_union' attribute is deprecated, use the
'unary_union()' method instead.
```

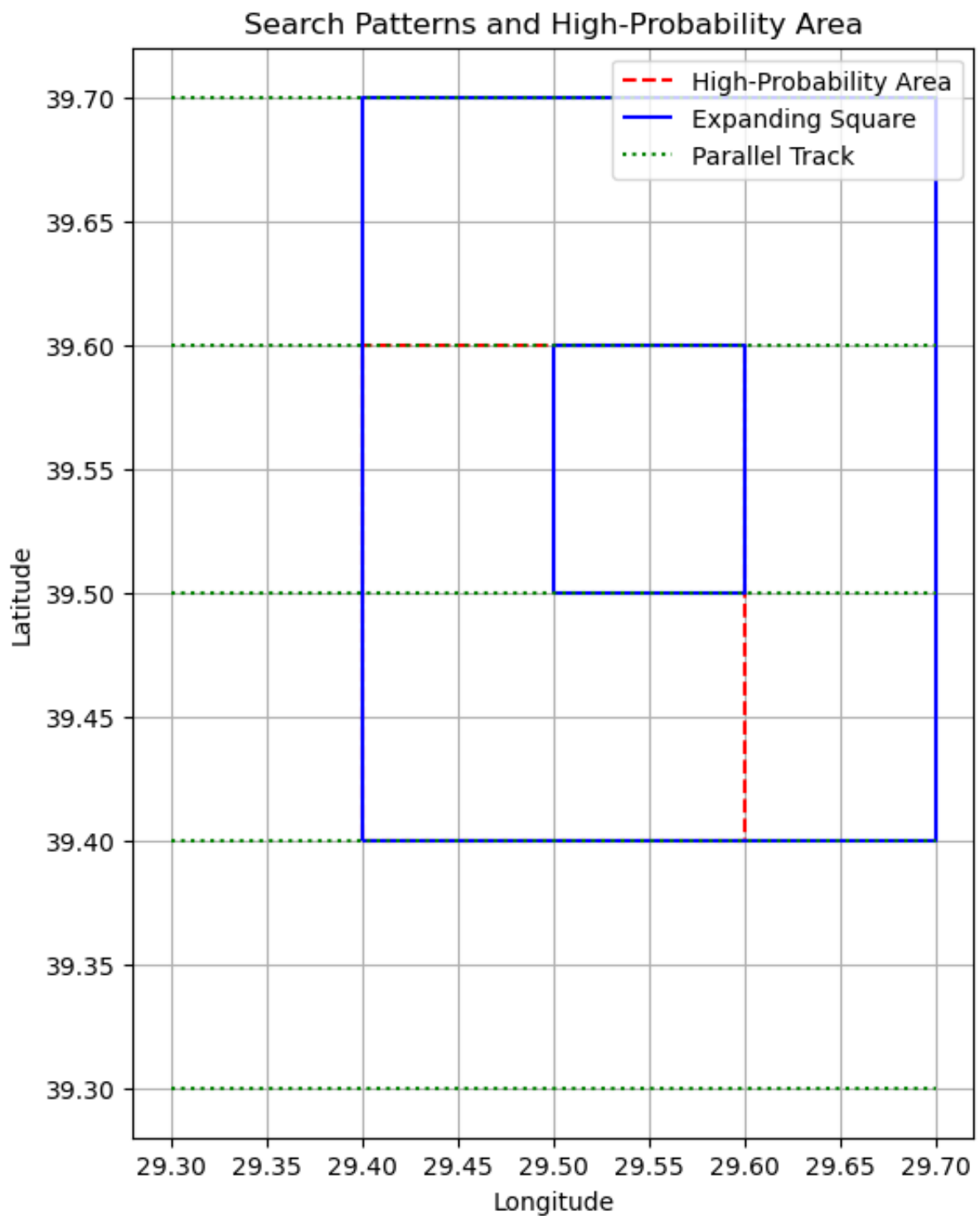
```
parallel_track_coverage = parallel_track_gdf.intersection(high_prob_area.u
nary_union).area.sum()
```

```
Expanding Square Coverage (area units): 0.0500000000000001425
Parallel Track Coverage (area units): 0.0
```

```
/var/folders/fp/drm5pw1d13x1n4wvd2q14v0m0000gn/T/ipykernel_26763/3804832417.  
py:3: UserWarning: Geometry is in a geographic CRS. Results from 'area' are  
likely incorrect. Use 'GeoSeries.to_crs()' to re-project geometries to a pro  
jected CRS before this operation.
```

```
parallel_track_coverage = parallel_track_gdf.intersection(high_prob_area.u  
nary_union).area.sum()
```

```
In [83]: # Visualize Search Patterns and High-Probability Area  
fig, ax = plt.subplots(figsize=(10, 8))  
  
high_prob_area.boundary.plot(ax=ax, color="red", linestyle="--", label="High  
expanding_square_gdf.boundary.plot(ax=ax, color="blue", label="Expanding Squ  
parallel_track_gdf.plot(ax=ax, color="green", linestyle=":", label="Parallel  
  
plt.title("Search Patterns and High-Probability Area")  
plt.xlabel("Longitude")  
plt.ylabel("Latitude")  
plt.legend()  
plt.grid()  
plt.show()
```



```
In [86]: # Streamlit UI Setup
st.title("Search Pattern Comparison and Probability Mapping")
st.sidebar.header("Settings")

# Step 1: Define the High-Probability Area
st.sidebar.subheader("High-Probability Area")
x_min = st.sidebar.slider("X Min", 29.0, 30.0, 29.4)
x_max = st.sidebar.slider("X Max", 29.0, 30.0, 29.6)
y_min = st.sidebar.slider("Y Min", 39.0, 40.0, 39.4)
y_max = st.sidebar.slider("Y Max", 39.0, 40.0, 39.6)

high_prob_area_coords = [(x_min, y_min), (x_max, y_min), (x_max, y_max), (x_
```

```

high_prob_area = gpd.GeoDataFrame(geometry=[Polygon(high_prob_area_coords)],

# Step 2: Generate Search Patterns
st.sidebar.subheader("Search Patterns")
pattern_type = st.sidebar.selectbox("Select Search Pattern", ["Expanding Squ

lkp = Point(29.5, 39.5)
search_area = Polygon([(29.3, 39.3), (29.7, 39.3), (29.7, 39.7), (29.3, 39.7

if pattern_type == "Expanding Square":
    expanding_square_coords = [
        [(29.5, 39.5), (29.6, 39.5), (29.6, 39.6), (29.5, 39.6), (29.5, 39.5
        [(29.4, 39.4), (29.7, 39.4), (29.7, 39.7), (29.4, 39.7), (29.4, 39.4
    ]
    search_pattern = gpd.GeoDataFrame(geometry=[Polygon(coords) for coords i

elif pattern_type == "Parallel Track":
    parallel_tracks = [
        LineString([(29.3, y), (29.7, y)]) for y in [39.3, 39.4, 39.5, 39.6,
    ]
    search_pattern = gpd.GeoDataFrame(geometry=parallel_tracks, crs="EPSG:43

# Step 3: Visualize Search Patterns and High-Probability Area
st.subheader("Search Pattern Visualization")
fig, ax = plt.subplots(figsize=(10, 8))

high_prob_area.boundary.plot(ax=ax, color="red", linestyle="--", label="High
search_pattern.boundary.plot(ax=ax, color="blue", label=pattern_type)
gpd.GeoSeries([lkp]).plot(ax=ax, color="black", label="LKP", marker="x")

ax.set_title(f"{pattern_type} Search Pattern and High-Probability Area")
ax.legend()
st.pyplot(fig)

# Step 4: Analyze Coverage
st.subheader("Coverage Analysis")
coverage = search_pattern.intersection(high_prob_area.unary_union).area.sum(
st.write(f"Coverage of the {pattern_type}: **{coverage:.2f} area units**")

# Step 5: Add Probability Map Visualization (Optional)
st.sidebar.subheader("Probability Map")
show_prob_map = st.sidebar.checkbox("Show Probability Map", value=False)

if show_prob_map:
    with rasterio.open("mcda_probability_map.tif") as src:
        prob_map = src.read(1)
        plt.figure(figsize=(10, 8))
        plt.imshow(prob_map, cmap="viridis", origin="upper")
        plt.colorbar(label="Probability")
        plt.title("Probability Map")
        plt.xlabel("Longitude")
        plt.ylabel("Latitude")
        st.pyplot()

```

2025-01-19 04:02:50.484 WARNING streamlit.runtime.scriptrunner\_utils.script\_run\_context: Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.534

**Warning:** to view this Streamlit app on a browser, run it with the following

command:

```
streamlit run /Users/lasyatummal/anaconda3/lib/python3.11/site-packages/ipykernel_launcher.py [ARGUMENTS]
```

2025-01-19 04:02:50.535 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.535 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.536 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.537 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.537 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.538 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.538 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.538 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.539 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.540 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.541 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.542 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.542 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.542 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.543 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.543 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.544 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.544 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

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2025-01-19 04:02:50.545 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

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2025-01-19 04:02:50.545 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.545 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

2025-01-19 04:02:50.546 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.

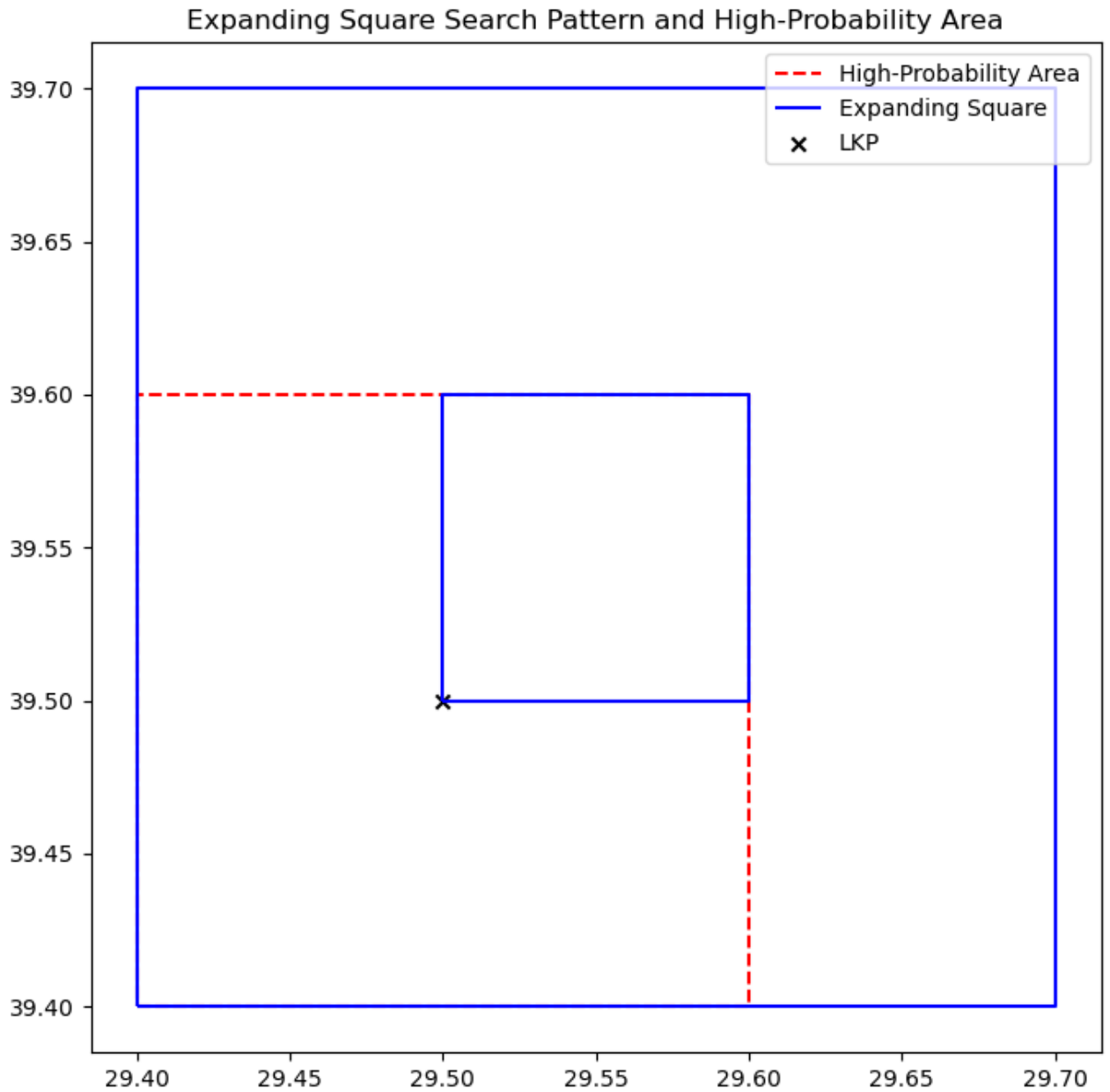
2025-01-19 04:02:50.546 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.



```
2025-01-19 04:02:50.575 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.576 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.577 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.577 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.578 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.579 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.580 Session state does not function when running a scrip
t without `streamlit run`
2025-01-19 04:02:50.581 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.581 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.584 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.585 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.765 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.919 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.919 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.920 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.920 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
/var/folders/fp/drm5pwld13xln4wvd2q14v0m0000gn/T/ipykernel_26763/539321407.p
y:49: DeprecationWarning: The 'unary_union' attribute is deprecated, use the
'union_all()' method instead.
    coverage = search_pattern.intersection(high_prob_area.unary_union).area.su
m()
/var/folders/fp/drm5pwld13xln4wvd2q14v0m0000gn/T/ipykernel_26763/539321407.p
y:49: UserWarning: Geometry is in a geographic CRS. Results from 'area' are
likely incorrect. Use 'GeoSeries.to_crs()' to re-project geometries to a pro
jected CRS before this operation.

    coverage = search_pattern.intersection(high_prob_area.unary_union).area.su
m()
2025-01-19 04:02:50.941 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.942 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.942 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.943 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.943 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.943 Thread 'MainThread': missing ScriptRunContext! This
warning can be ignored when running in bare mode.
2025-01-19 04:02:50.944 Thread 'MainThread': missing ScriptRunContext! This
```

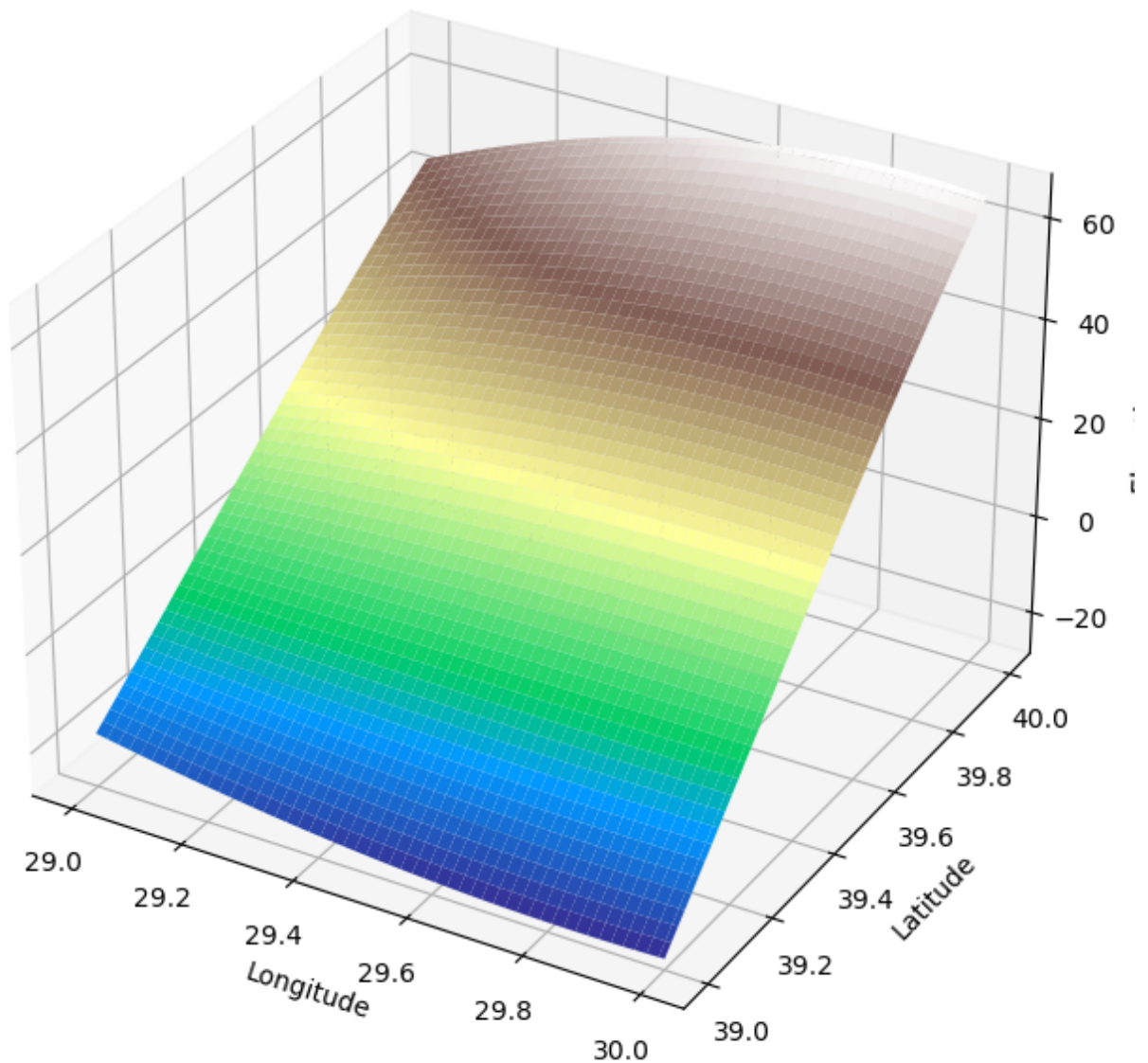
warning can be ignored when running in bare mode.  
2025-01-19 04:02:50.944 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.  
2025-01-19 04:02:50.945 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.  
2025-01-19 04:02:50.946 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.  
2025-01-19 04:02:50.946 Thread 'MainThread': missing ScriptRunContext! This warning can be ignored when running in bare mode.



```
In [88]: # Simulated elevation data for a search grid
x = np.linspace(29.0, 30.0, 100) # Longitude
y = np.linspace(39.0, 40.0, 100) # Latitude
xv, yv = np.meshgrid(x, y)
elevation = np.sin(xv) * np.cos(yv) * 100 # Simulated terrain elevation

# Plot the 3D terrain
fig = plt.figure(figsize=(10, 8))
ax = fig.add_subplot(111, projection="3d")
ax.plot_surface(xv, yv, elevation, cmap="terrain", edgecolor="none")
ax.set_title("3D Terrain for Search Area")
ax.set_xlabel("Longitude")
ax.set_ylabel("Latitude")
ax.set_zlabel("Elevation (m)")
plt.show()
```

3D Terrain for Search Area



In [ ]:

