

# capstone\_code

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## 1 Plot the Likelihood of Finding Cobalt Deposits in British Columbia

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### 1.0.1 Load packages and data

```
[1]: # Load packages using correct env
import geopandas as gpd
import shapely.geometry as shp
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches
import seaborn as sns
import numpy as np
import random
%matplotlib inline

# Import data
INPUT_FILE = "BedrockP.shp"
data = gpd.read_file(INPUT_FILE)
```

### 1.0.2 Preview data structure, info, and columns

```
[2]: data.head(2)
```

```
[2]:
```

	gid	upid	area_m2	strat_unit	era	period	\
0	33649.0	9c192ffb	4368828.0	uKPCv1	Mesozoic	Cretaceous	
1	26344.0	824df15e	34431394.0	1KGsv	Mesozoic	Cretaceous	

	strat_age	strat_name	gp_suite	\
0	Upper Cretaceous	Powell Creek Formation - upper unit	None	
1	Lower Cretaceous	Gambier Group	Gambier Group	

	fm_lithodm	...	terrane	basin	basin_age	project	\
0	Powell Creek Formation	...	Overlap	None	None	Chilcotin-Bonaparte	
1	None	...	Overlap	None	None	Mid-coast	

```

src_url \
0 http://www.em.gov.bc.ca/Mining/Geoscience/Publ...
1 http://www.em.gov.bc.ca/Mining/Geoscience/Publ...

src_ref_s \
0 Schiarizza et al., 1997, Chilcotin-Bonaparte, ...
1 Bellefontaine et al., 1994, Mid-coast, BCGS, 0...

map_comp edit_date \
0 Schiarizza, 2017, Chicotin-Bonaparte 2018-04-05
1 Massey et al., 2005, BC 2018-04-05

pub_org \
0 British Columbia Geological Survey
1 British Columbia Geological Survey

geometry
0 POLYGON ((466212.8026486822 5669781.23969778, ...
1 POLYGON ((541326.0118804163 5529535.058946755,...

[2 rows x 28 columns]
```

```
[3]: data.info()
```

```

<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 1596 entries, 0 to 1595
Data columns (total 28 columns):
gid          1596 non-null float64
upid         1596 non-null object
area_m2      1596 non-null float64
strat_unit   1596 non-null object
era          1596 non-null object
period       1596 non-null object
strat_age    1596 non-null object
strat_name   1184 non-null object
gp_suite     1036 non-null object
fm_lithodm   443 non-null object
mem_phase    76 non-null object
rock_class   1596 non-null object
rock_type    1596 non-null object
rk_char      2 non-null object
unit_desc    1596 non-null object
age_max      1596 non-null object
age_min      1596 non-null object
belt         1596 non-null object
terrane      1596 non-null object
basin        240 non-null object
```

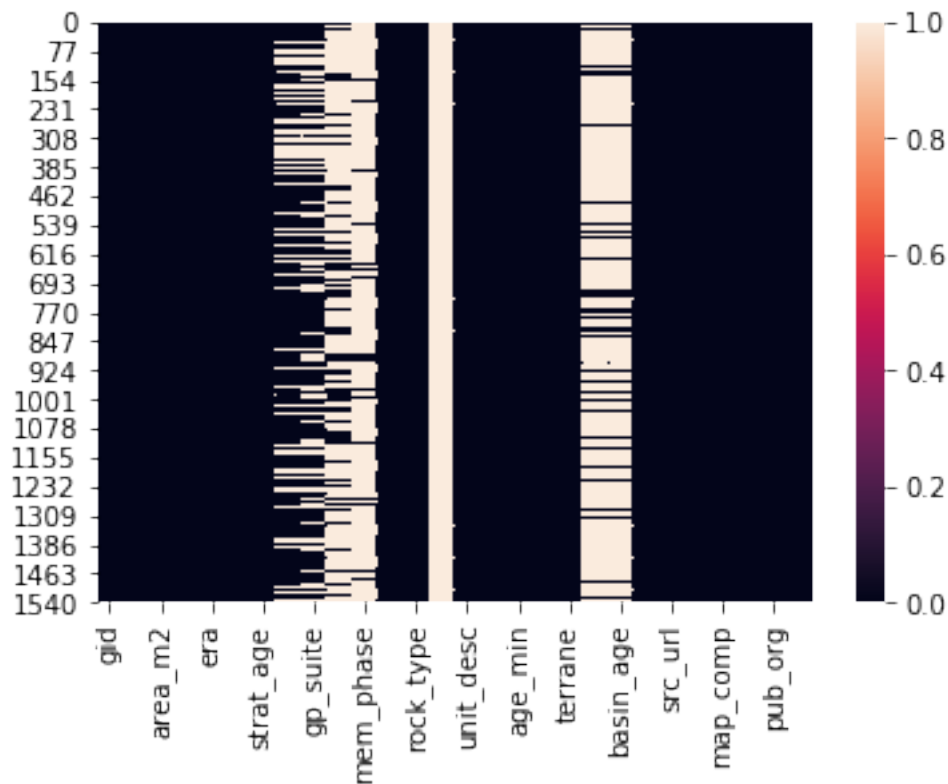
```

basin_age      240 non-null object
project        1596 non-null object
src_url        1596 non-null object
src_ref_s      1596 non-null object
map_comp       1596 non-null object
edit_date      1596 non-null object
pub_org        1596 non-null object
geometry        1596 non-null object
dtypes: float64(2), object(26)
memory usage: 349.2+ KB

```

```
[4]: # View any missing data
sns.heatmap(data.isna())
```

```
[4]: <matplotlib.axes._subplots.AxesSubplot at 0x7f953a2055c0>
```



These missing data seem irrelevant to the end goal of creating heat map, so ignore for now

```
[5]: # View some columns with granodioritic and ultramafic data
data.iloc[12:19,4:13]
```

```

[5]:
era                                period \
12 Paleozoic-Mesozoic Carboniferous to Jurassic
13      Mesozoic      Cretaceous
14      Cenozoic      Paleogene
15      Cenozoic      Paleogene
16 Paleozoic-Mesozoic Carboniferous to Jurassic
17      Paleozoic      Permian
18      Mesozoic      Jurassic to Cretaceous

strat_age \
12 Middle Mississippian to Middle Jurassic
13      Lower Cretaceous
14      Eocene
15      Eocene
16 Middle Mississippian to Middle Jurassic
17      Permian
18      Late Jurassic to Early Cretaceous

strat_name \
12      Bridge River Complex
13      Taylor Creek Group - Lizard Formation
14      Mission Ridge suite
15      Mission Ridge suite
16      Bridge River Complex
17 Shulaps Ultramafic Complex - Serpentinite Mela...
18      None

gp_suite      fm_lithodm mem_phase \
12      Bridge River Complex      None      None
13      Taylor Creek Group      Lizard Formation      None
14      Mission Ridge suite      None      None
15      Mission Ridge suite      None      None
16      Bridge River Complex      None      None
17 Shulaps Ultramafic Complex      Serpentinite Melange unit      None
18      None      None      None

rock_class      rock_type
12 sedimentary rocks      marine sedimentary and volcanic rocks
13 sedimentary rocks      undivided sedimentary rocks
14      intrusive rocks      granodioritic intrusive rocks
15      intrusive rocks      granodioritic intrusive rocks
16 sedimentary rocks      marine sedimentary and volcanic rocks
17      ultramafic rocks      ultramafic rocks
18      intrusive rocks      quartz dioritic intrusive rocks

```

View the number of rock classes and types

```
[6]: # Rock class counts
data['rock_class'].value_counts()
```

```
[6]: sedimentary rocks      556
      intrusive rocks      455
      volcanic rocks       435
      metamorphic rocks    104
      ultramafic rocks      46
      Name: rock_class, dtype: int64
```

```
[7]: # Unique rock types
nrocks = data['rock_type'].nunique()
print(f"Number of rock_types: {nrocks}")

# View most common rock types and where gran/serp/ultra line up
data['rock_type'].value_counts().head(12)
```

Number of rock\_types: 43

```
[7]: undivided sedimentary rocks      246
      undivided volcanic rocks      149
      granodioritic intrusive rocks  147
      marine sedimentary and volcanic rocks 146
      basaltic volcanic rocks      141
      dioritic intrusive rocks      96
      coarse clastic sedimentary rocks 79
      quartz dioritic intrusive rocks 72
      ultramafic rocks              46
      calc-alkaline volcanic rocks    44
      serpentinite ultramafic rocks   42
      high level quartz phyric, felsitic intrusive rocks 39
      Name: rock_type, dtype: int64
```

### 1.0.3 Create labels for the two relevant rock\_types (gran + serp/ultra)

```
[8]: # Label for 1 granodioritic rock_type
allrocks = list(data['rock_type'].unique())
rocks = [x for x in allrocks if 'grano' in x]

# Labels for both ultramafic rock types
rocks.extend([x for x in allrocks if 'ultramafic' in x])
print(rocks)
```

```
['granodioritic intrusive rocks', 'ultramafic rocks', 'serpentinite ultramafic rocks']
```

```
[9]: # Get indexes of 2 major rock type labels
gran_idx = data['rock_type'].apply(lambda x: x == rocks[0])
serp_idx = data['rock_type'].apply(lambda x: x in rocks[1:])
all_idx = (gran_idx | serp_idx)

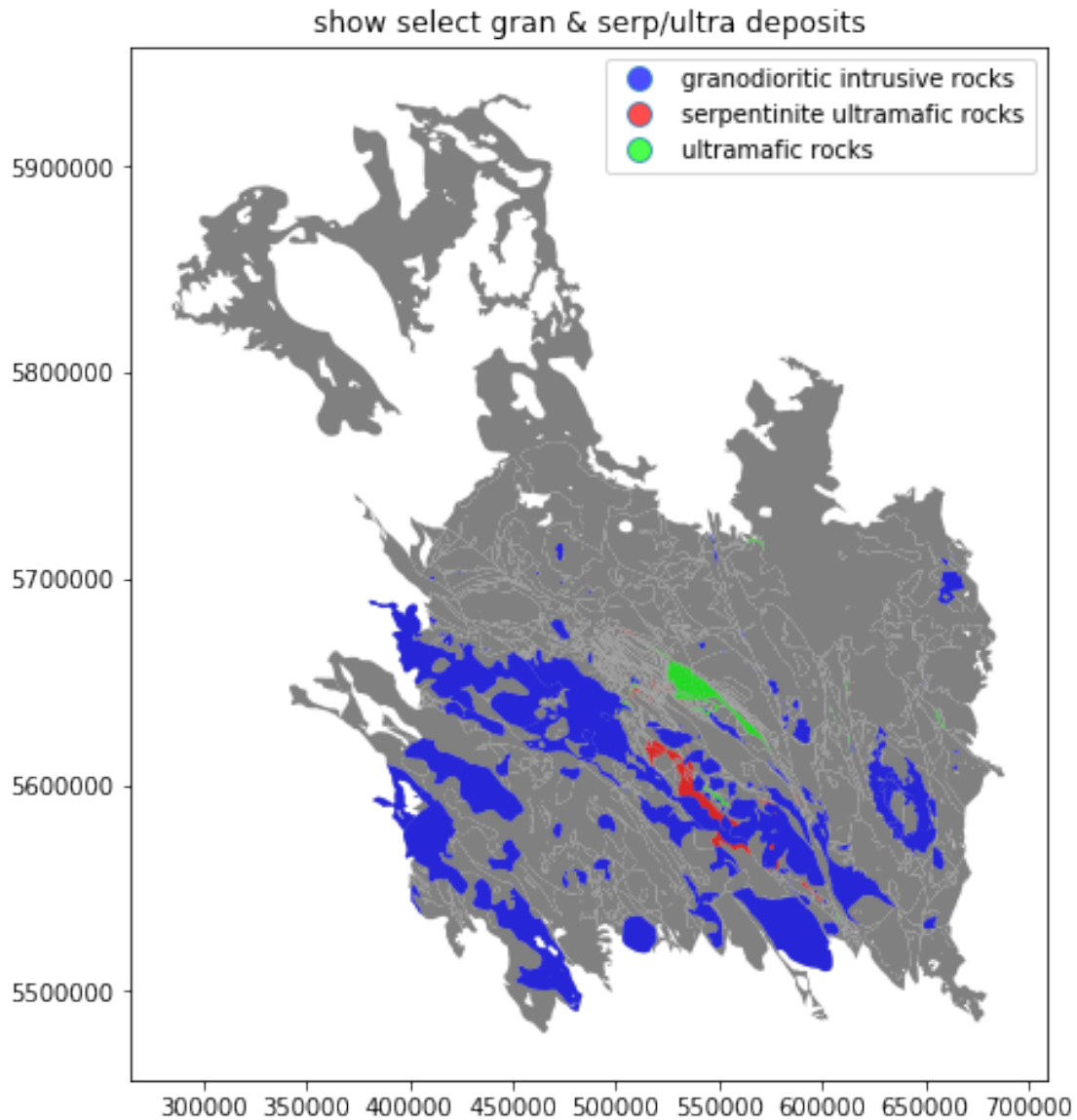
# Verify that sum matches value_counts() above: 147 (gran), 88 (46 ultra + 42
↳serp)
print(f"Total gran deposits: {sum(gran_idx)}\nTotal serp deposits:
↳{sum(serp_idx)}")
```

Total gran deposits: 147

Total serp deposits: 88

#### 1.0.4 Plot only deposits with relevant rocks

```
[10]: fig, ax = plt.subplots(1, 1, figsize=(8,8))
data.plot(ax=ax, color='grey')
data[all_idx].plot(ax=ax, column="rock_type", legend=True, cmap='brg', alpha=0.
↳7)
ax.title.set_text('show select gran & serp/ultra deposits')
```



Although ultramafic and serpentinite labels are shown individually here, for the rest of the analysis, they will be combined (see instructions)

### 1.0.5 View coordinate reference system (crs) of data set

EPSG: 26910 (NAD83/UTM Zone 10N).

```
[11]: print(data.crs)
```

```
{'init': 'epsg:26910'}
```

### 1.0.6 Calculate likelihood of finding a cobalt deposit for all regions

Find intersecting regions at a distance  $r$  from relevant deposits, then assign a likelihood of finding cobalt in that region. The instructions state that the likelihood should be maximum along where granodiorite and serpentite regions touch, then fall smoothly to zero at a distance  $R = 10 \text{ km}$ . A likelihood function that matches these goals can be described by the function

$$L(r) = 1 - r/R. \quad (1)$$

The max distance  $R$  determines when the likelihood should go to zero when  $r \geq R$ .

```
[12]: # Set buffer resolution, buffer distances (in meters), & blank heatdf
buffer_num = 1+10
buffer_max = 10000
distances = np.linspace(0, buffer_max, buffer_num)
heatdf = gpd.GeoDataFrame()

# Loop over buffer regions to find intersections & set equal to same likelihood
for dist in distances:
    # Enlarge regions by a buffer distance
    granbuffer = data.loc[gran_idx, 'geometry'].buffer(dist)
    serpbuffer = data.loc[serp_idx, 'geometry'].buffer(dist)

    # Merge all buffers into MultiPolygon, then recast as GeoDataFrame
    grantotal = gpd.GeoDataFrame({'geometry':granbuffer.unary_union, 'rock':
    ↪'gran'})
    serptotal = gpd.GeoDataFrame({'geometry':serpbuffer.unary_union, 'rock':
    ↪'serp'})

    # Get all intersecting regions between gran/serp at this dist & assign
    ↪likelihood
    inter = gpd.overlay(grantotal, serptotal, how='intersection')
    inter['likelihood'] = 1 - dist/buffer_max

    # Store results in heatdf
    heatdf = heatdf.append(inter)

# Post-process heatdf to improve plotting & preserve crs
heatdf = heatdf.sort_values(by='likelihood', axis=0)
heatdf = heatdf.reset_index()
heatdf = heatdf.drop(labels=['index', 'rock_1', 'rock_2'], axis=1)
heatdf.crs = {'init': 'epsg:26910'}

[13]: # Select a random point to show likelihood at in region (if turned on)
heatdf['coords'] = heatdf['geometry'].boundary.apply(lambda x: x.
    ↪representative_point().coords[:])
heatdf['coords'] = [random.choice(coords) for coords in heatdf['coords']]
```



### 1.0.7 Plot heat maps

The full heat maps show that the majority of the interfacial regions between granodioritic and the serpentite/ultramafic rocks, hence, the likelihood, are in the lower portion of the plot. I will plot both the full heat map, as well as a zoomed in version to see more details.

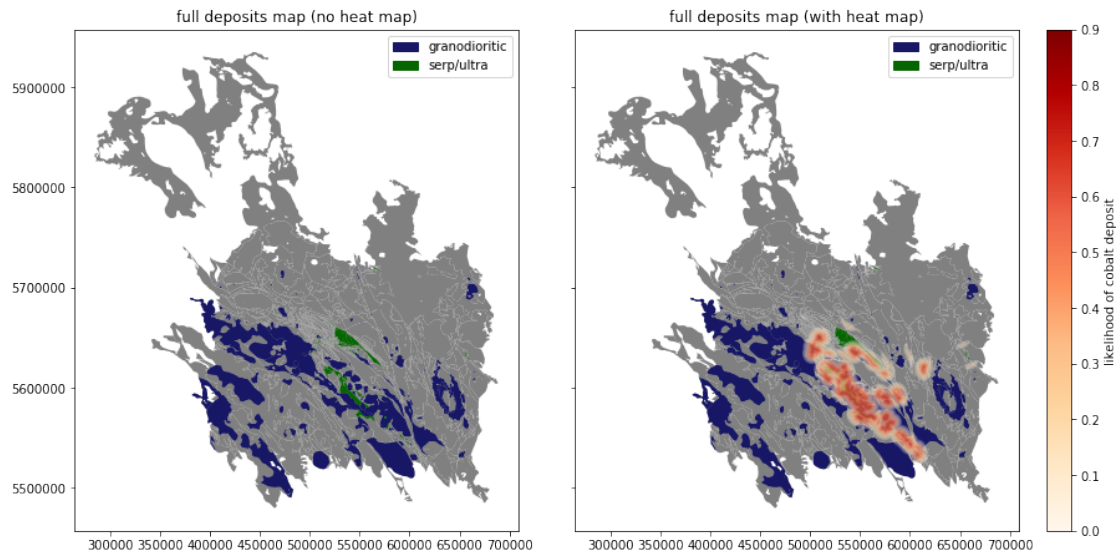
```
[14]: # Create plot variables: zoom axis limits, RGB colors, legend handles/text
xlm = [490000, 620000]
ylm = [5520000, 5680000]
gran_color = (0.09, 0.09, 0.40)    # dark blue
serp_color = (0.02, 0.40, 0.00)    # dark green
gran_patch = mpatches.Patch(color=gran_color)
serp_patch = mpatches.Patch(color=serp_color)
handles = [gran_patch, serp_patch]
leg = ['granodioritic', 'serp/ultra']

# Full deposits plot (no colorbar)
fig, axs = plt.subplots(1, 2, sharex=True, sharey=True, figsize=(16,7.4))
data.plot(ax=axs[0], color='grey')
data[gran_idx].plot(ax=axs[0], color=gran_color)
data[serp_idx].plot(ax=axs[0], color=serp_color)
axs[0].title.set_text('full deposits map (no heat map)')
axs[0].legend(handles=handles, labels=leg, loc='upper right')

# Full deposits plot (with colorbar)
data.plot(ax=axs[1], color='grey')
data[gran_idx].plot(ax=axs[1], color=gran_color)
data[serp_idx].plot(ax=axs[1], color=serp_color)
heatdf.plot(ax=axs[1], column='likelihood', alpha=0.2, cmap='OrRd')
axs[1].set_title('full deposits map (with heat map)');
axs[1].legend(handles=handles, labels=leg, loc='upper right')
plt.subplots_adjust(wspace=0.0)

# Create custom colorbar
cim = plt.cm.ScalarMappable(cmap='OrRd',
                           norm=plt.Normalize(vmin=heatdf['likelihood'].min(),
                                                vmax=heatdf['likelihood'].max()))
cbar = plt.colorbar(cim, orientation='vertical')
cbar.set_label('likelihood of cobalt deposit')
plt.subplots_adjust(wspace=0.0)

# Save figure (keep commented until ready). figsizes (20,9), (16,7.4), (13,6)
# plt.savefig('full-small.png')
```



```
[15]: # Zoomed deposits plot (no colorbar)
fig, axs = plt.subplots(1, 2, sharex=True, sharey=True, figsize=(16,8))
data.plot(ax=axs[0], color='grey')
data[gran_idx].plot(ax=axs[0], color=gran_color)
data[serp_idx].plot(ax=axs[0], color=serp_color)
axs[0].title.set_text('zoomed rock deposits (no heat map)')
axs[0].legend(handles=handles, labels=leg, loc='upper right')
axs[0].set_xlim(xlm);
axs[0].set_ylim(ylm);

# Zoomed deposits plot (with colorbar)
data.plot(ax=axs[1], color='grey')
data[gran_idx].plot(ax=axs[1], color=gran_color)
data[serp_idx].plot(ax=axs[1], color=serp_color)
heatdf.plot(ax=axs[1], column='likelihood', alpha=0.25, cmap='OrRd',
            edgecolor='black')
axs[1].title.set_text('zoomed rock deposits (with heat map)')
axs[1].legend(handles=handles, labels=leg, loc='upper right', fancybox=True)
axs[1].set_xlim(xlm);
axs[1].set_ylim(ylm);

# Annotate plot with a subset of likelihoods spaced out from each other
ptlist = []
for idx, row in heatdf.iterrows():
    lhood = row['likelihood']
    cpt = shp.Point(row['coords'])
    check = sum([cpt.within(pt.buffer(7000)) for pt in ptlist])
```

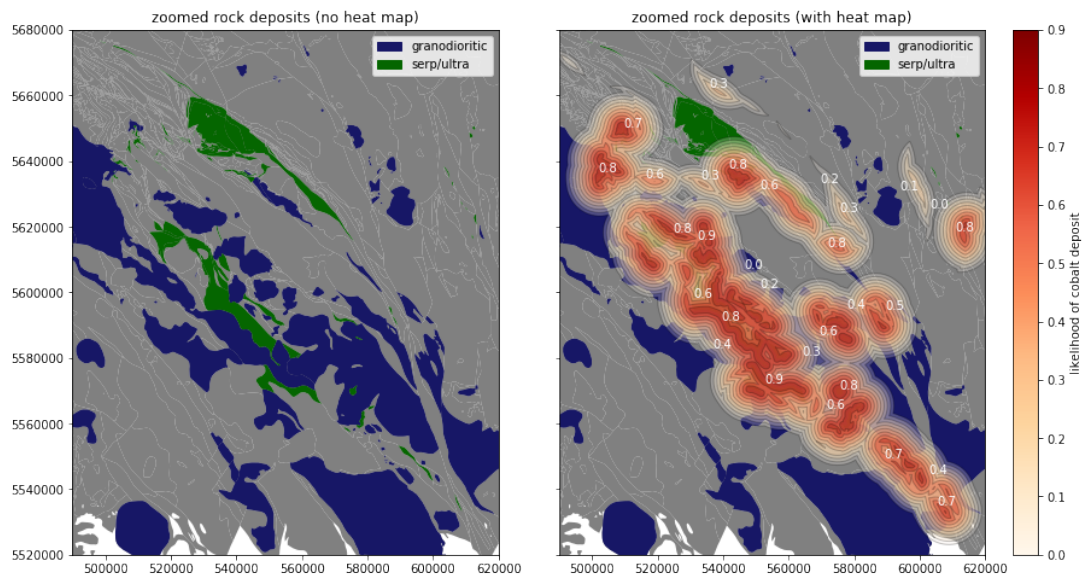
```

# Check if coord is not within specified buffer of any point in ptlist
if check == 0:
    ptlist.append(cpt)
    label = round(lhood, 2)
    plt.annotate(s=label, xy=row['coords'], ha='center', color='white')

# Create custom colorbar
cim = plt.cm.ScalarMappable(cmap='OrRd',
                           norm=plt.Normalize(vmin=heatdf['likelihood'].min(),
                                                vmax=heatdf['likelihood'].max()))
cbar = plt.colorbar(cim, orientation='vertical')
cbar.set_label('likelihood of cobalt deposit')
plt.subplots_adjust(wspace=0.0)

# Save figure (comment until ready). figsizes large:(20,10), med:(16,8), small:
→ (13,6)
# plt.savefig('zoomed-med.png')

```



### 1.0.8 Create a function to query exact likelihoods at any coord

The map provides a good way to quickly see general areas of high likelihood, but displaying exact likelihoods at every location cluttered the map too much. So if an exact coordinate is provided, the function below will return the numerical likelihood at that location.

```
[16]: def likelihood(coord):
    """ Given a coord tuple of (x, y), print the numerical likelihood of
    ↪ finding cobalt
    coord (540000, 5660000) should yield l=0.3
    coord (600000, 5540000) should yield l=0.7
    coord (578000, 5560000) should yield l=0.9
    coord (520000, 5540000) should yield l=0.0"""

    cpt = shp.Point(coord)
    ptwithinlist = [cpt.within(shape) for shape in heatdf['geometry']]
    hits = sum(ptwithinlist)

    if hits == 0:
        maxhit = 0.0
    else:
        maxhit = max(heatdf['likelihood'][ptwithinlist])

    print(f"The likelihood of finding cobalt at {coord} = {round(maxhit, 2)}")
```

```
[17]: # Test cases
likelihood((540000, 5660000))
likelihood((600000, 5540000))
likelihood((578000, 5560000))
likelihood((520000, 5540000))
```

```
The likelihood of finding cobalt at (540000, 5660000) = 0.3
The likelihood of finding cobalt at (600000, 5540000) = 0.7
The likelihood of finding cobalt at (578000, 5560000) = 0.9
The likelihood of finding cobalt at (520000, 5540000) = 0.0
```