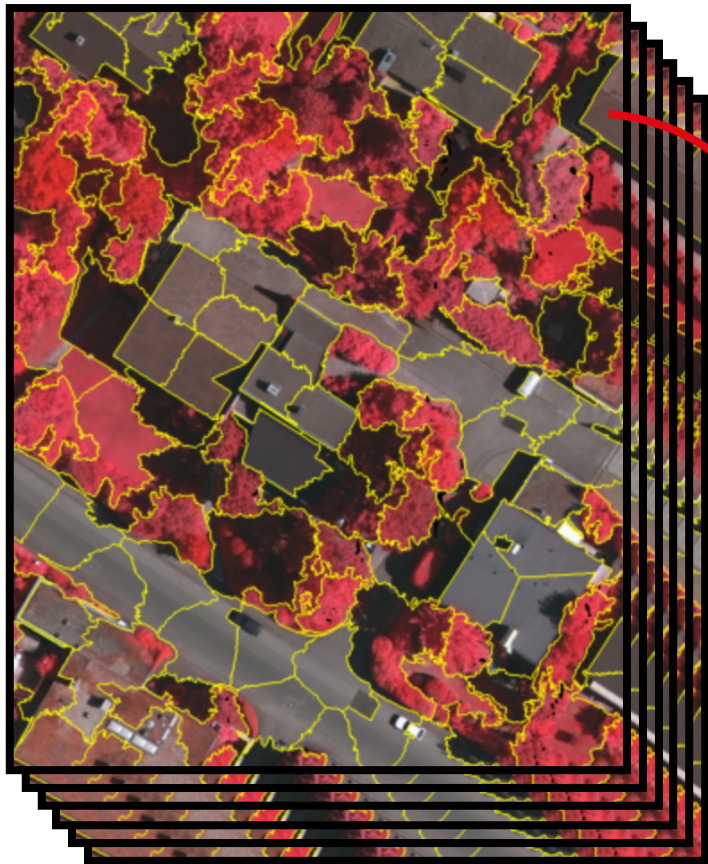


The background is a collage of various images: a topographic map with a river, a satellite view of a city, a snowy mountain, a forest, and a grassy field with small objects highlighted by bounding boxes. A large red rectangle is positioned on the right side, containing the title.

Image classification - Part 1

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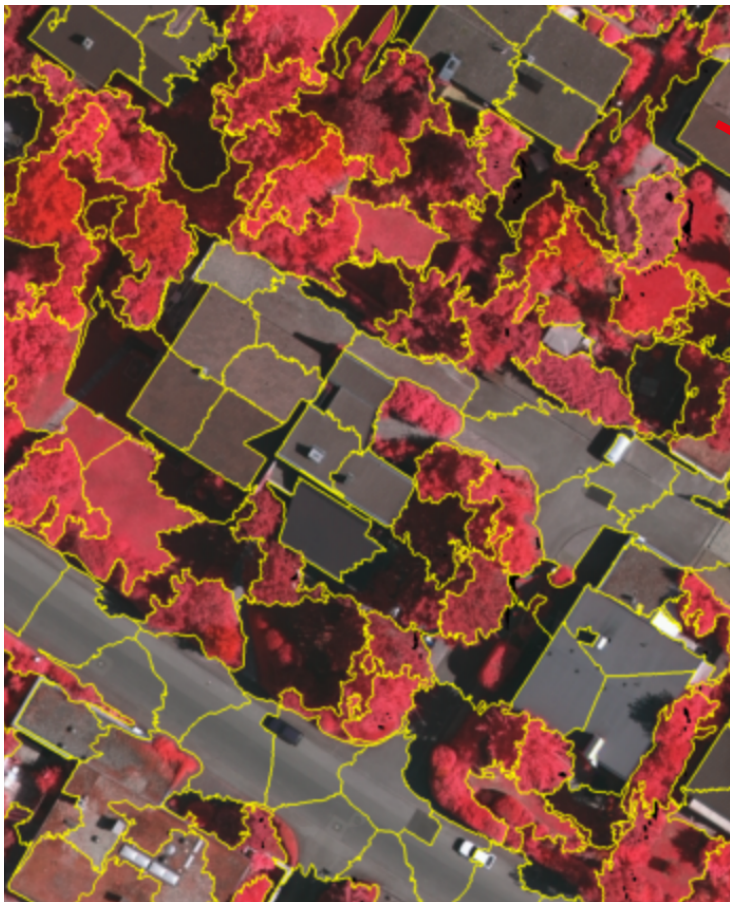


Features

$$\mathbf{x}_i = \left[\mathbf{x}_i^{\text{av}} \quad \mathbf{x}_i^{\text{std}} \quad \mathbf{x}_i^{\text{hist}} \right]$$

EPFL Train set: Image features and targets for each region

3

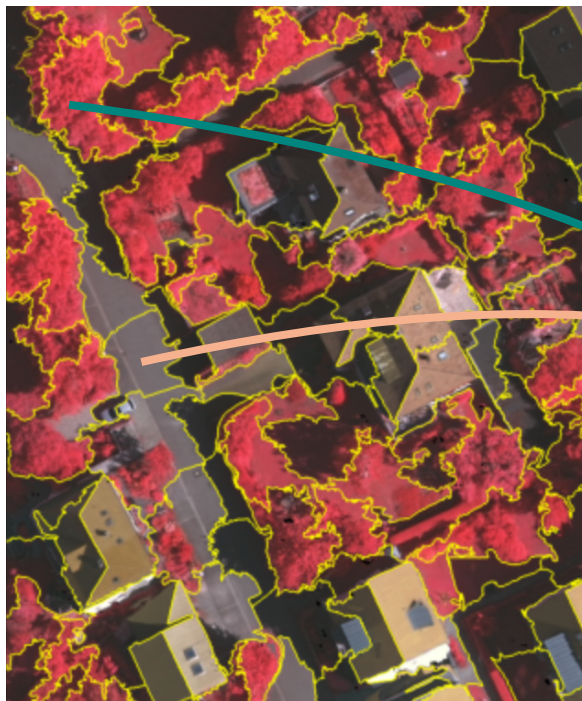


Features

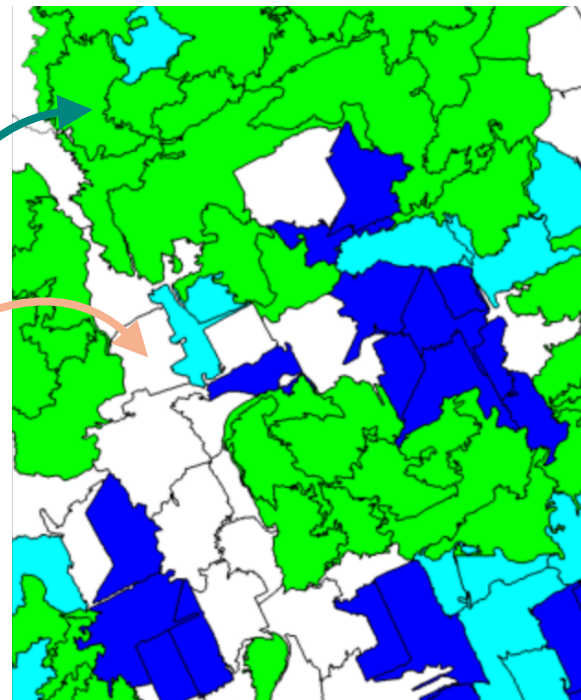
$$\mathbf{x}_i = \left[\mathbf{x}_i^{\text{av}} \quad \mathbf{x}_i^{\text{std}} \quad \mathbf{x}_i^{\text{hist}} \right]$$

Ground Truth / Reference / Target

- 0: Impervious
- 1 : Building
- 2: Lower vegetation
- 3: Tree



Machine
Learning
Model



- Impervious
- Building
- Low vegetation
- Tree

```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
# Load data example
iris_dataset = load_iris()
features = iris_dataset["data"]
labels = iris_dataset["target"]
# Get features and targets of the train and test sets
train_features, test_features, train_labels, test_labels = train_test_split(features,
                                                                              labels,
                                                                              test_size=0.33,
                                                                              random_state=42)
```

```
# Shape of train features (2D array)
print(train_features.shape)
```

(100, 4)

```
# Shape of train labels (1D array)
print(train_labels.shape)
```

(100,)

Training a classifier

```
from sklearn.ensemble import RandomForestClassifier

# Normalize features
mean_array = np.mean(train_features, axis=0)
std_array = np.std(train_features, axis=0)
norm_train_features = (train_features - mean_array) / std_array

# Create random forest classifier
classifier = RandomForestClassifier(random_state=10)

# Fit model
classifier.fit(norm_train_features, train_labels)
```

Prediction

```
# Normalize features
norm_test_features = (test_features - mean_array) / std_array

# Predict labels for the test set
label_predictions = classifier.predict(norm_test_features)
```

- Read the provided PDF file and Jupyter Notebook with detailed instructions
- Tasks:
 1. Install dependencies, if needed
 2. Extract regions and their features
 3. Create a training dataset, with the following classes
 - 0: Impervious
 - 1: Building
 - 2: Low vegetation
 - 3: Tree
 4. Normalize features
 5. Train Random Forest classifier
 6. Predict classification maps
 7. Visualize predictions
 8. Answer the questions of the PDF file of instructions

Concatenate arrays



```
import numpy as np

array_1 = np.array([10, 70, 60])
array_2 = np.array([50, 150, 200])

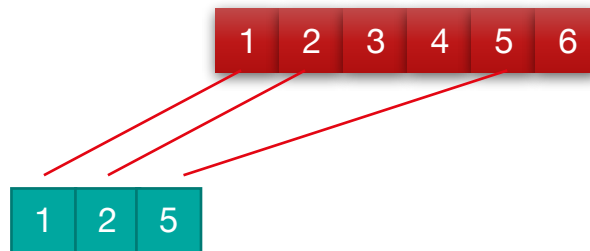
list_of_arrays = [array_1, array_2]
print(list_of_arrays)

[array([10, 70, 60]), array([ 50, 150, 200])]

merged_array = np.concatenate(list_of_arrays)
print(merged_array)

[ 10  70  60  50 150 200]
```

randomly split train test



```
train_image_numbers = np.random.choice([1, 2, 3, 4, 5, 6],
                                       size=3,
                                       replace=False)
```



```
test_image_numbers = []
for i in all_image_numbers:
    if i not in train_image_numbers:
        test_image_numbers.append(i)
```

```
# alternative one liner
test_image_numbers = [i for i in all_image_numbers
                      if i not in train_image_numbers]
```

Change Values at selected locations

Image

50	90
120	100

region ids

4	5
5	5

mask

false	true
true	true

pixels of region 5

90	120	100
----	-----	-----

changed image

50	255
255	255

```
image = np.array([[ 50, 90],  
                  [120, 100]])
```

```
regions = np.array([[4, 5],  
                    [5, 5]])
```

```
mask_region_5 = regions == 5  
print(mask_region_5)
```

```
[[False  True]  
 [ True  True]]
```

```
# Select pixels of the region 5  
print(image[mask_region_5])
```

```
[ 90 120 100]
```

```
# Change value of pixels of the region 5  
image[mask_region_5] = 255
```

```
print(image)
```

```
[[ 50 255]  
 [255 255]]
```

EPFL Find label for region

which label should
region 5 have?

0 or 1?

region ids

4	5
5	5

label map

0	0
1	1

mask region

false	true
true	true

mask label 0

true	true
false	false

label * region

0	1
0	0

sum → 1 pixel

mask label 1

false	false
true	true

0	0
1	1

sum → 2 pixels

for each label_id
↓

label 0

label 1

```
regions = np.array([[4, 5],
                    [5, 5]])
```

```
label_map = np.array([[0, 0],
                     [1, 1]])
```

Computing the best label for region 5

```
mask_region_5 = regions == 5
```

```
num_labels = 2
```

```
intersection_per_label = []
```

```
for label_id in range(num_labels):
```

```
    mask_label = label_map == label_id
```

Compute intersection of each region with each label

```
    intersection = np.sum(mask_region_5 * mask_label)
```

```
    intersection_per_label.append(intersection)
```

```
intersection_per_label = np.array(intersection_per_label)
```

```
print(intersection_per_label)
```

```
[1 2]
```

Obtain the index of the label with largest intersection

```
selected_label = np.argmax(intersection_per_label)
```

```
print(selected_label)
```

```
1
```