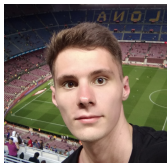
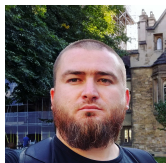


# Deep Learning for classification and segmentation for understanding clouds from satellite images

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# The problem

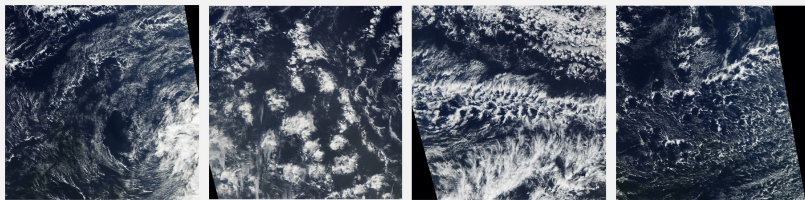
**Task :** With a given satellite image, we predict segmentation masks for four cloud organization patterns : **Sugar**, **Flower**, **Fish**, and **Gravel**.

**Motivation :** Shallow clouds play a large role in the Earth's radiation balance and yet they are poorly represented in **climate models**.

**Dataset :** Gathered by the scientists of two meteorologic institutions and later was used to host a Data Science **competition on Kaggle**.

**Problematics :** The ground-truth masks presented in the **dataset** are quite **noisy** meaning they include a lot of areas that actually do not contain clouds at all. Also, the masks of different classes can overlap. These two facts significantly increase **problem complexity**.

**Metric :** **Mean Dice coefficient** is chosen as the metric to measure the model performance.



(a) Sugar

(b) Flower

(c) Fish

(d) Gravel

**FIGURE** – Canonical examples of the four cloud organization patterns.

# Research

## Several aspects for studying :

- How training on **different image sizes** affects the final results ?
- What is an **optimal threshold** value for pseudo-labels generation ?
- Which **encoder** should be used with **U-Net** architecture ?

## What we propose :

- **Two-stage segmentation** pipeline with **progressive resizing** & **pseudo-labelling**.
- **Post-processing** procedure after the model generates predictions, namely, drop all the masks which **pixel size**<sup>1</sup> is less than some **parameter  $k$**  (can be optimized).

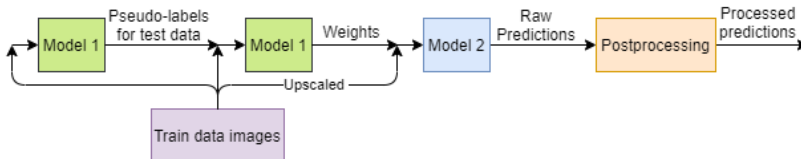


FIGURE – Pipeline.

1. Pixel size is defined as a number of pixels which belong to the same connected component

# Summary

## Results :

- Dice score :  $0.658^2$ .
- Position : Top 6.2% [95th].
- Final leaderbord ranking by Dice coefficient differs by no more than  $\sim 1,5\%$  between the 1st and the 100th positions.

## Ways to improve :

- Specific pre-processing which tackles the problem of noisy dataset.
- Training on bigger image sizes.
- Better post-processing threshold selection.