



St Petersburg  
University



BIO  
GEO  
HAB

# Underwater biotope mapping: automatic processing of underwater video data

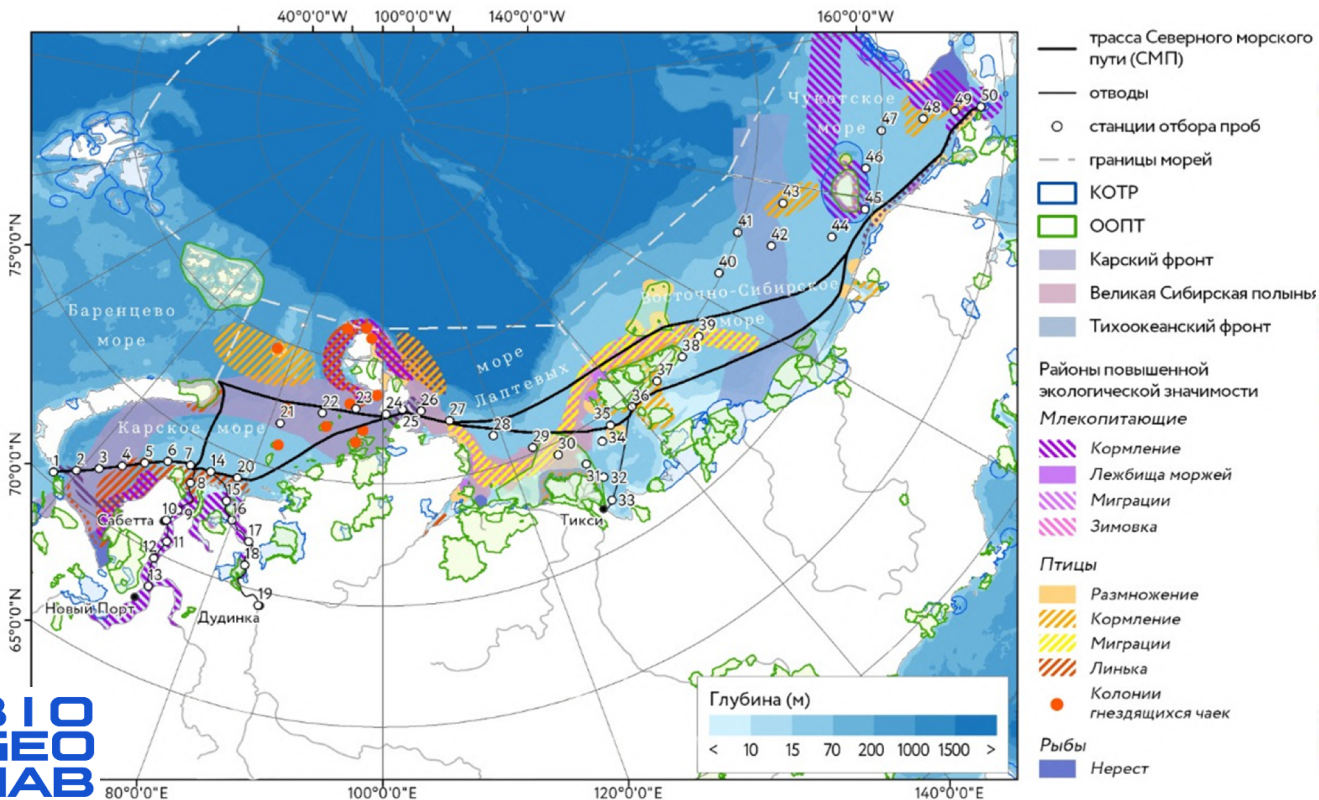
Oleg Iakushkin, Ekaterina Pavlova, Anastasiya Iavrova, Evgeniy Pen, Anna Frikh-Khar, Yana Terekhina, Anna Bulanova, Nikolay Shabalin, and Olga Sedova



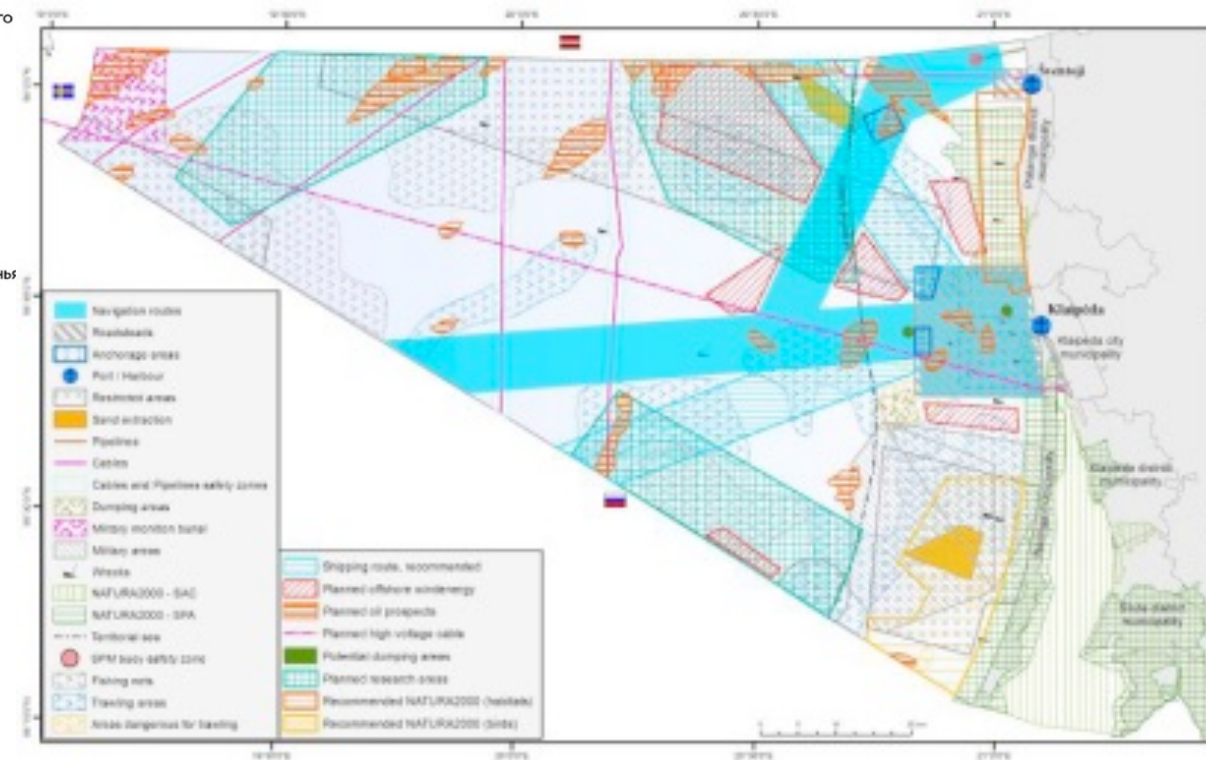
# Why scan the sea?

1. Fish biomes are heavily correlated with sea bottom
2. Construction projects require sea bed surveys for permissions and compensations
3. Biomes on the sea floor tend to correlate with minerals and deposits under the sea bed

## Construction and operation of structures on the Arctic shelf of the Russian Federation



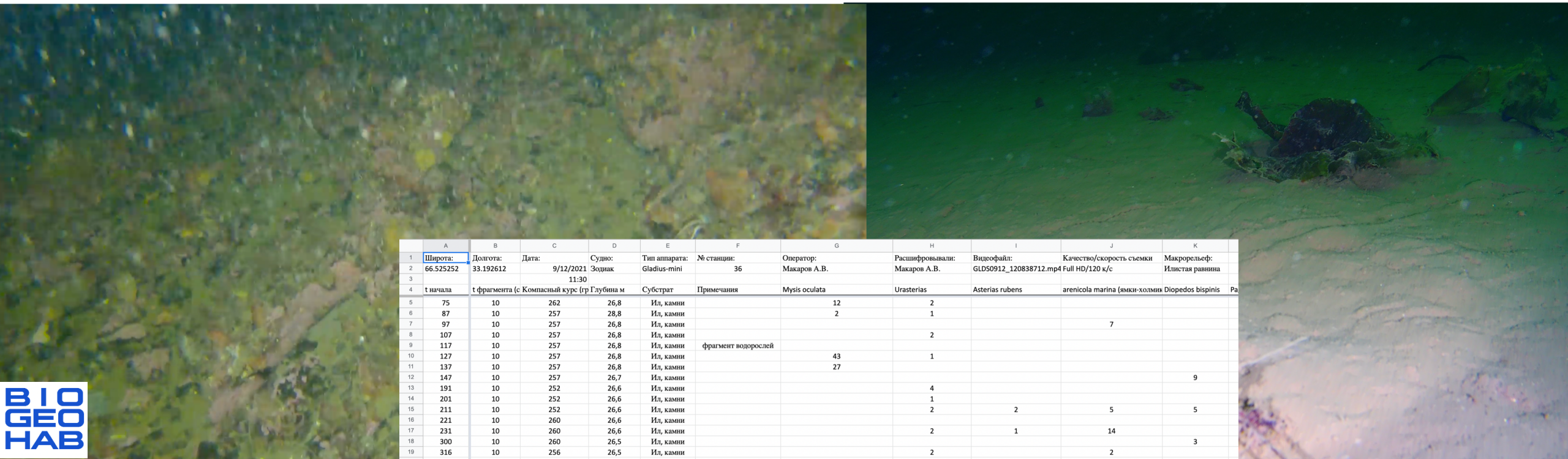
## Marine Spatial Planning



# Problem definition

## It takes lots of time to turn a video into a table

- A specially trained scientists looks thru the video and classifies plants and animals
- It takes approximately **2 hours** for him to **process 5 minute video**
- A scientist has specialized training that allows him to see not only images but also integrate context



|    | A         | B              | C                  | D         | E             | F                   | G             | H               | I                      | J                              | K                 |    |
|----|-----------|----------------|--------------------|-----------|---------------|---------------------|---------------|-----------------|------------------------|--------------------------------|-------------------|----|
| 1  | Широта:   | Долгота:       | Дата:              | Судно:    | Тип аппарата: | № станции:          | Оператор:     | Расшифровывали: | Видеофайл:             | Качество/скорость съемки       | Макрорельеф:      |    |
| 2  | 66.525252 | 33.192612      | 9/12/2021          | Зодиак    | Gladius-mini  | 36                  | Макаров А.В.  | Макаров А.В.    | GLD50912_120838712.mp4 | Full HD/120 к/с                | Илистая равнина   |    |
| 3  |           |                | 11:30              |           |               |                     |               |                 |                        |                                |                   |    |
| 4  | t начала  | t фрагмента (с | Компасный курс (гр | Глубина м | Субстрат      | Примечания          | Mysis oculata | Urasterias      | Asterias rubens        | arenicola marina (ямки-холмики | Diopados bispinis | Pa |
| 5  | 75        | 10             | 262                | 26,8      | Ил, камни     |                     | 12            | 2               |                        |                                |                   |    |
| 6  | 87        | 10             | 257                | 28,8      | Ил, камни     |                     | 2             | 1               |                        |                                |                   |    |
| 7  | 97        | 10             | 257                | 26,8      | Ил, камни     |                     |               |                 |                        | 7                              |                   |    |
| 8  | 107       | 10             | 257                | 26,8      | Ил, камни     |                     |               | 2               |                        |                                |                   |    |
| 9  | 117       | 10             | 257                | 26,8      | Ил, камни     | фрагмент водорослей |               |                 |                        |                                |                   |    |
| 10 | 127       | 10             | 257                | 26,8      | Ил, камни     |                     | 43            | 1               |                        |                                |                   |    |
| 11 | 137       | 10             | 257                | 26,8      | Ил, камни     |                     | 27            |                 |                        |                                |                   |    |
| 12 | 147       | 10             | 257                | 26,7      | Ил, камни     |                     |               |                 |                        |                                | 9                 |    |
| 13 | 191       | 10             | 252                | 26,6      | Ил, камни     |                     |               | 4               |                        |                                |                   |    |
| 14 | 201       | 10             | 252                | 26,6      | Ил, камни     |                     |               | 1               |                        |                                |                   |    |
| 15 | 211       | 10             | 252                | 26,6      | Ил, камни     |                     |               | 2               | 2                      | 5                              | 5                 |    |
| 16 | 221       | 10             | 260                | 26,6      | Ил, камни     |                     |               |                 |                        |                                |                   |    |
| 17 | 231       | 10             | 260                | 26,6      | Ил, камни     |                     |               | 2               | 1                      | 14                             |                   |    |
| 18 | 300       | 10             | 260                | 26,5      | Ил, камни     |                     |               |                 |                        |                                | 3                 |    |
| 19 | 316       | 10             | 256                | 26,5      | Ил, камни     |                     |               | 2               |                        | 2                              |                   |    |







## Problem definition

- Videos are blurry chunky and low resolution
- There is plankton, dust and excrement's flying all around
- Animals have a wide size differential



Temperature: 11.3°C

Depth: -16.3m



## Our dataset and ResNet training

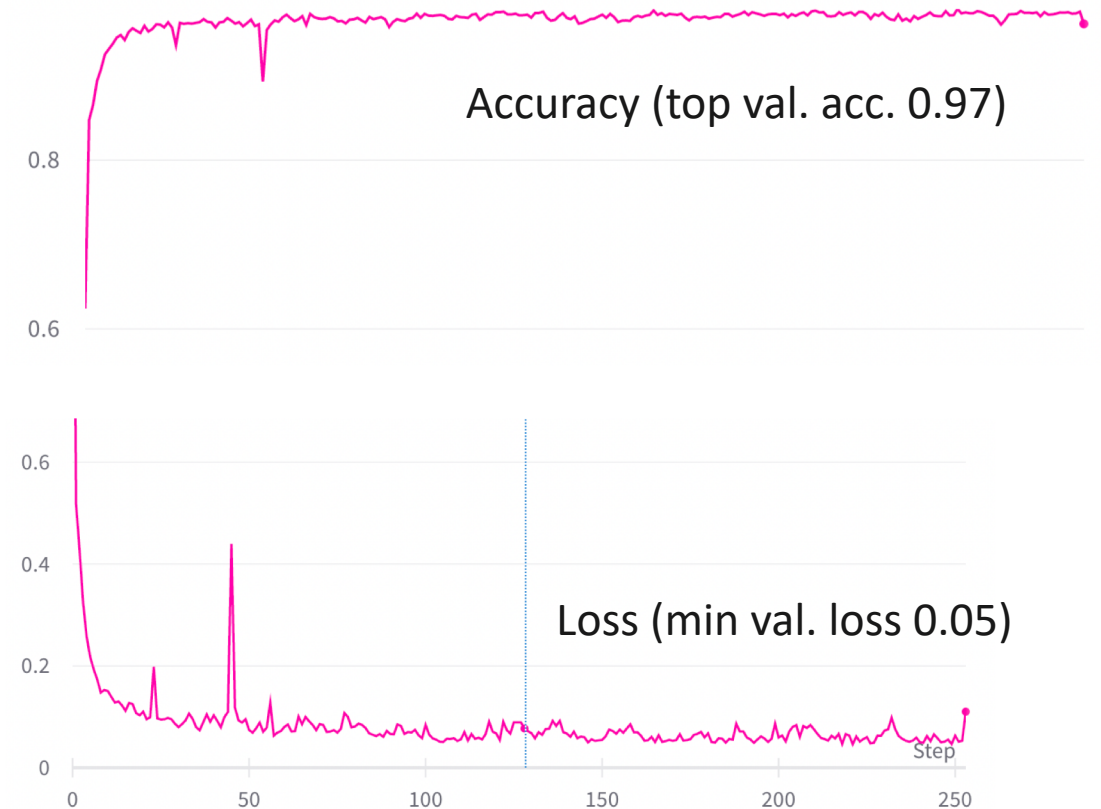
- **36 classes** special for for White Sea (30 animals, trash, none, human, fish, 2 types of plants)
- **60\40 mix** of chunks from video to internet aggregated images





# Our dataset and ResNet training

- During training all is scaled down to **250x250**
- Training is performed **from pretrained ResNet50** on 1k classes, all layer involved (**2h on A30**)
- Augmentations performed  
 Resize((250, 250)),  
 AutoAugment(),  
 RandomHorizontalFlip(),  
 RandomVerticalFlip(),  
 Normalize([0.485, 0.456, 0.406], [0.229, 0.224, 0.225]),  
 RandomErasing(),  
 Lambda(*lambda* x : x + torch.randn\_like(x)\*0.2)



## Our solution

1. Remove blurry frames, take only key frames
2. Split frames into chunks
3. Perform ResNet50 trained on our dataset on each chunk
4. Count animals presence
5. Generate table

---

Verify and sign generated table on expert-user side




## Our solution deployment – WASM





1. Prototype – python, Collaboratory
  2. WASM compiled – Runs in a user Web Browser
- FFmpeg (WASM compiled)
  - Piodite (bundled with OpenCV) detecting blurry images, cutting chunks
  - Onnx-runtime (PyTorch Resnet compiled and deployed in WASM thru it)
  - Exceljs (results table generation)





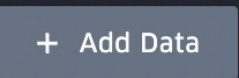
St Petersburg  
University


Current work in progress


 **kepler.gl**  
2.5.5









Datasets (2) 

 new dataset >  
54,936 rows




 GLDS0727\_153200794.csv >  
27 rows

Layers 




Ascidia  
Point






none  
Point

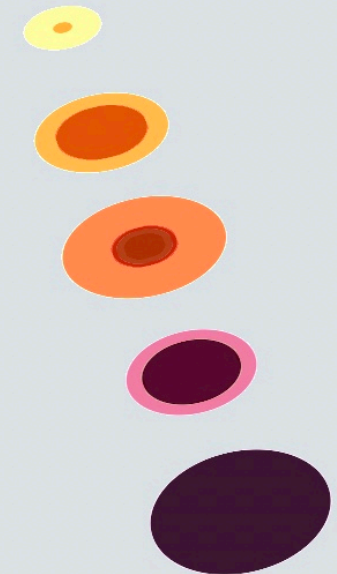


arenicola  
Point



Cnidaria  
Point







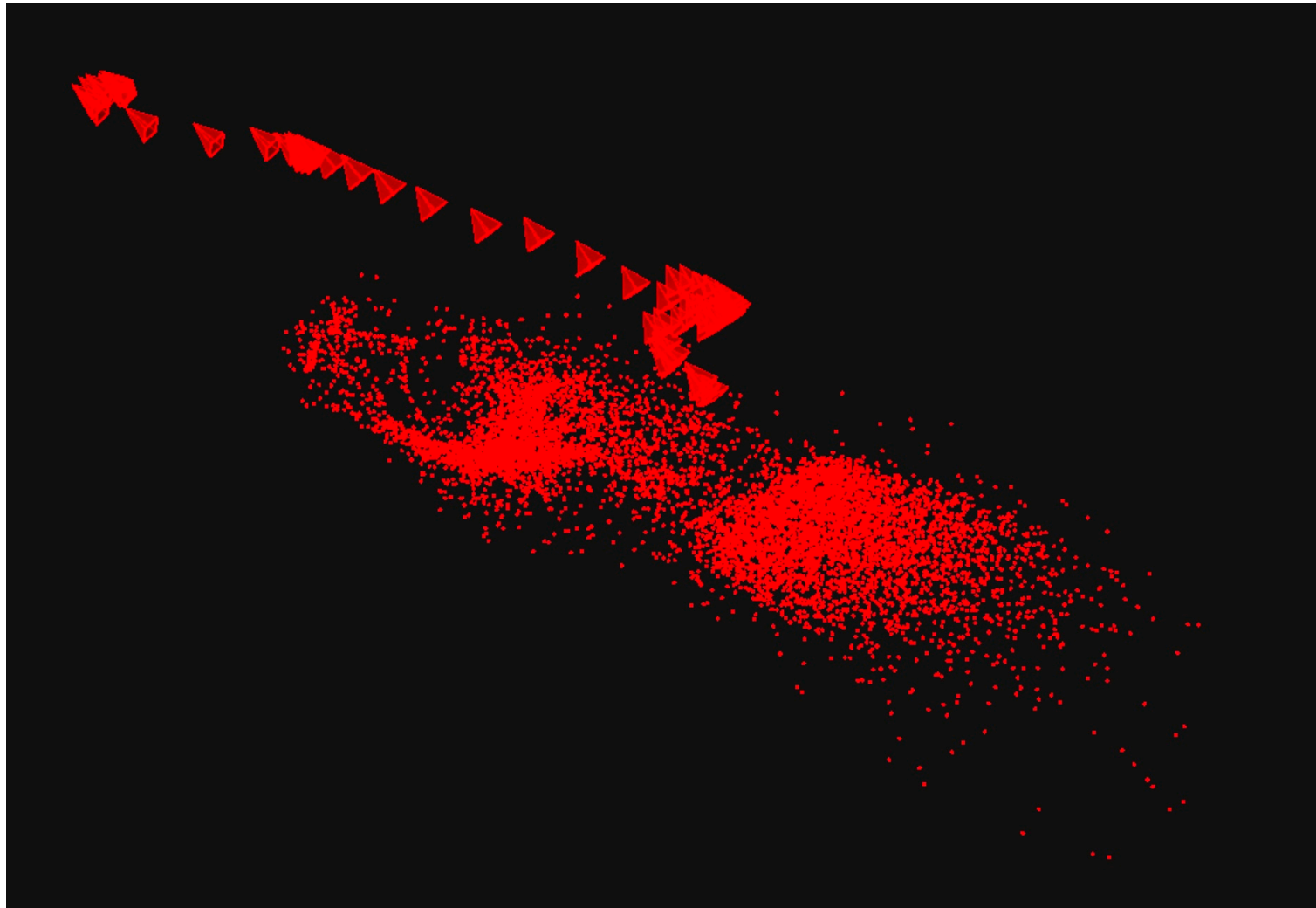
# How to count animals correctly and estimate volume? (part 1)

- FSL on a creature type
  - 5 class images required
  - ResNet detection helps to estimate where to apply FSL



## How to count animals correctly and estimate volume? (part 2)

- Video SLAM (HLOC)
- SfM (SPSR)
- Segmentation Projection
- Relative volume estimation







St Petersburg  
University

*Future is so interesting!*