



Cell nuclei segmentation: support vector machine

*Project proposal by
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Second Milestone

Implement Support Vector machine

First Milestone

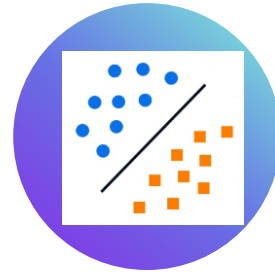
Implement Dice score

Third Milestone

Implement Pre-processing

Our data

Timeline



• Our data

○ 28 images of nuclei

○ *N2DH-GOWT1*

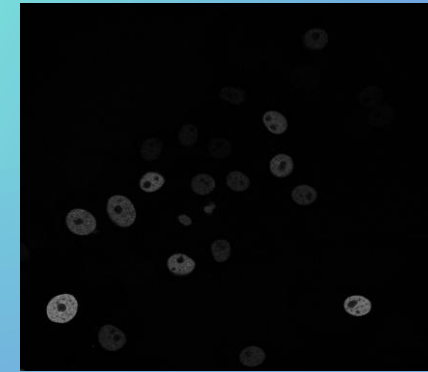
GFP transfected GOWT1 mouse embryonic stem cells

○ *N2DL-HeLa*

Histone 2B (H2B)-GFP expressing HeLa cells

○ *NIH3T3*

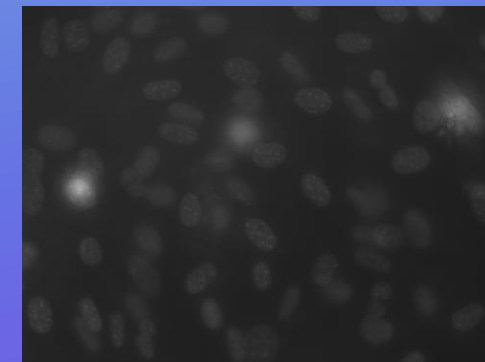
mouse embryonic fibroblast – CD tagged (EGFP)



N2DH-GOWT1



N2DL-HeLa



NIH3T3



Our data

N2DH-GOWT1

- Medium difficulty
- Heterogeneous staining
- Prominent nucleoli + mitoses
- Cells entering and leaving the field of view and frequent cell collisions

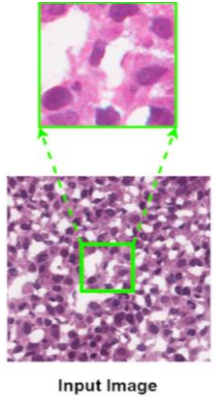
N2DL-HeLa

- High difficulty
- High cell density and low resolution
- Frequent mitoses
- Presence of colliding, entering and leaving cells with low fluorescence intensity

NIH3D3

- Medium difficulty
- Nuclei far apart + less clustering
- Nuclei homogeneous in shape and size
- But they also vary in brightness between images
- Images often contain visible debris.

Our goal

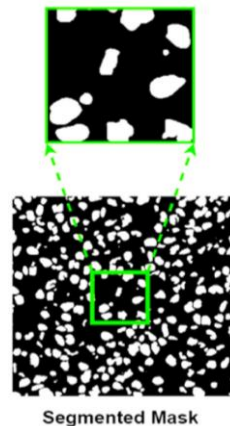


Input: Microscopic images

Preprocessing & filtering

Segmentation:
Support vector machine

Data mining: Counting nuclei

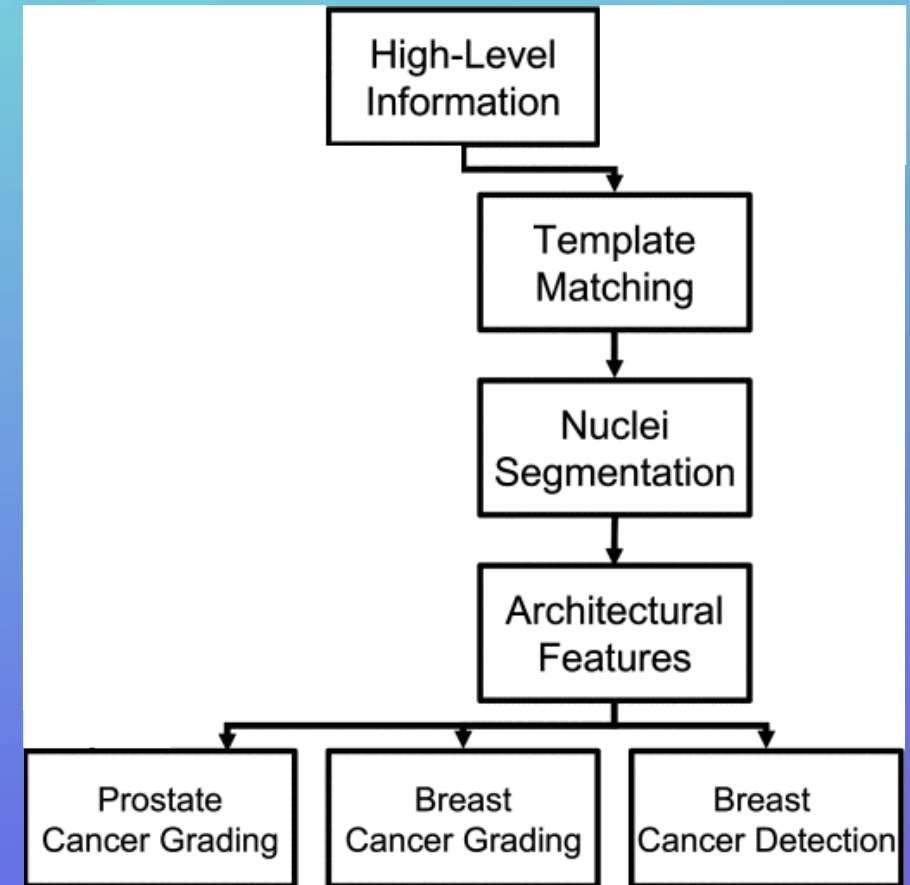


Output: Segmented image & number of nuclei

Our goal

Cell nuclei segmentation using support vector machine

Quantification of cell nuclei



quantify degree of malignancy (= grading)
→ key feature to predict patient prognosis
and decide on treatment options

Second Milestone

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Third Milestone

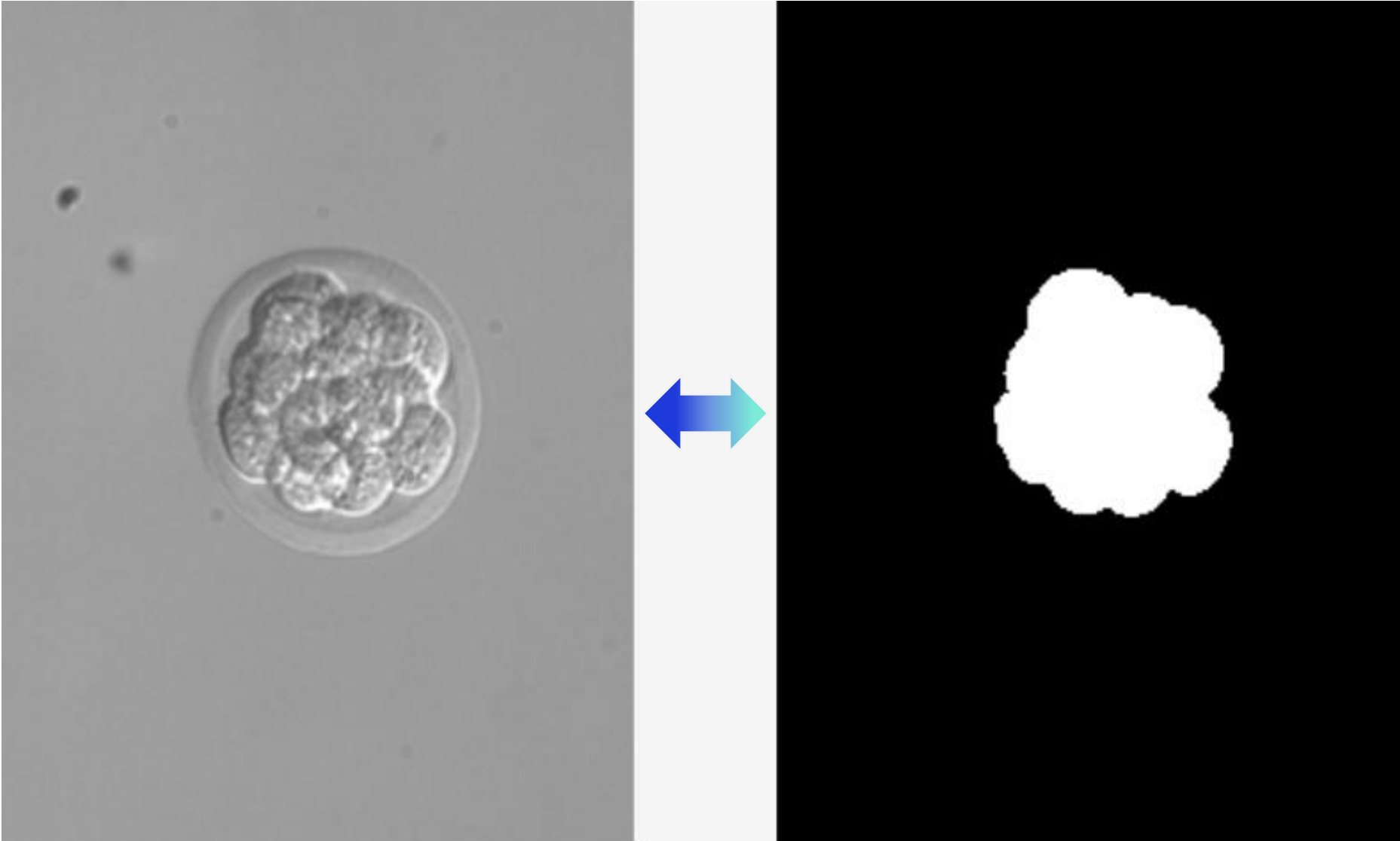
Implement Pre-processing

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Timeline



Dice Score



Dice Score

Evaluating the quality of the segmentation quantitatively

The part segmented by the algorithm

Object to segment

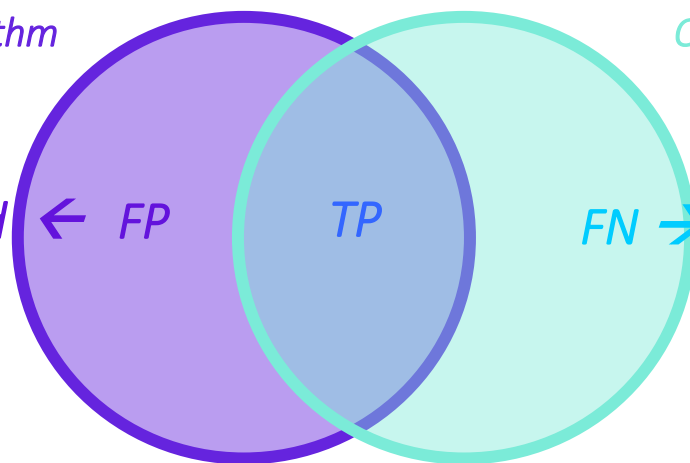
falsely detected

$\leftarrow FP$

TP

$\rightarrow FN$

not detected



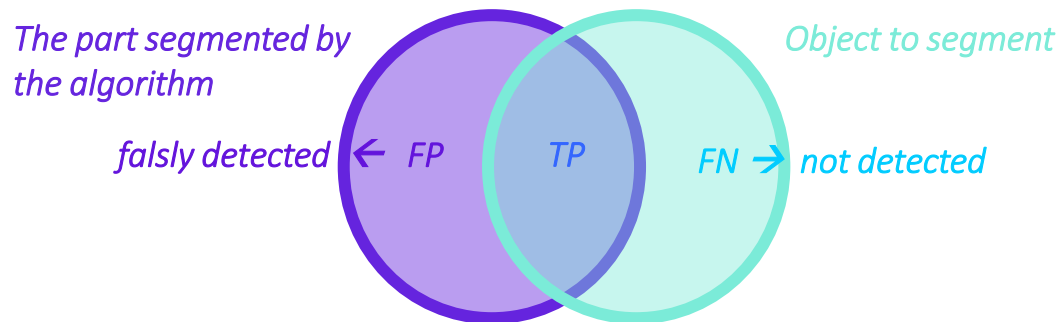
—————• $IoU = \frac{Intersection}{Union} = \frac{TP}{TP+FP+FN}$ \leftarrow not as well differentiable

—————• Better: $Dice = \frac{2*Intersection}{Union+Intersection} = F_1 = \frac{1}{\frac{1}{Prec} + \frac{1}{Recall}} = \frac{2 TP}{2 TP+FP+FN}$

—————• $Loss = 1 - Dice \in [0;1]$

—————• $Loss = 1 - \frac{2 \sum bf}{\sum b+f}$

Dice Score



- $$\text{IoU} = \frac{\text{Intersection}}{\text{Union}} = \frac{TP}{TP+FP+FN} \quad \leftarrow \text{not as well differentiable}$$

- Better:
$$\text{Dice} = \frac{2 * \text{Intersection}}{\text{Union} + \text{Intersection}} = F_1 = \frac{1}{\frac{1}{\text{Prec}} + \frac{1}{\text{Recall}}} = \frac{2 TP}{2 TP + FP + FN}$$

- $$\text{Loss} = 1 - \text{Dice} \in [0;1]$$

- $$\text{Loss} = 1 - \frac{2 \sum bf}{\sum b + f}$$

Planned analysis steps

- Write code for Dice-Score function
- Unit-testing
- Write code for synthetic images

Characteristics of the first Milestone

- Measure for evaluating our model

The purpose

- Quantify the quality of our model

Second Milestone

Implement Support Vector machine

First Milestone

Implement Dice score

Third Milestone

Implement Pre-processing

Our data

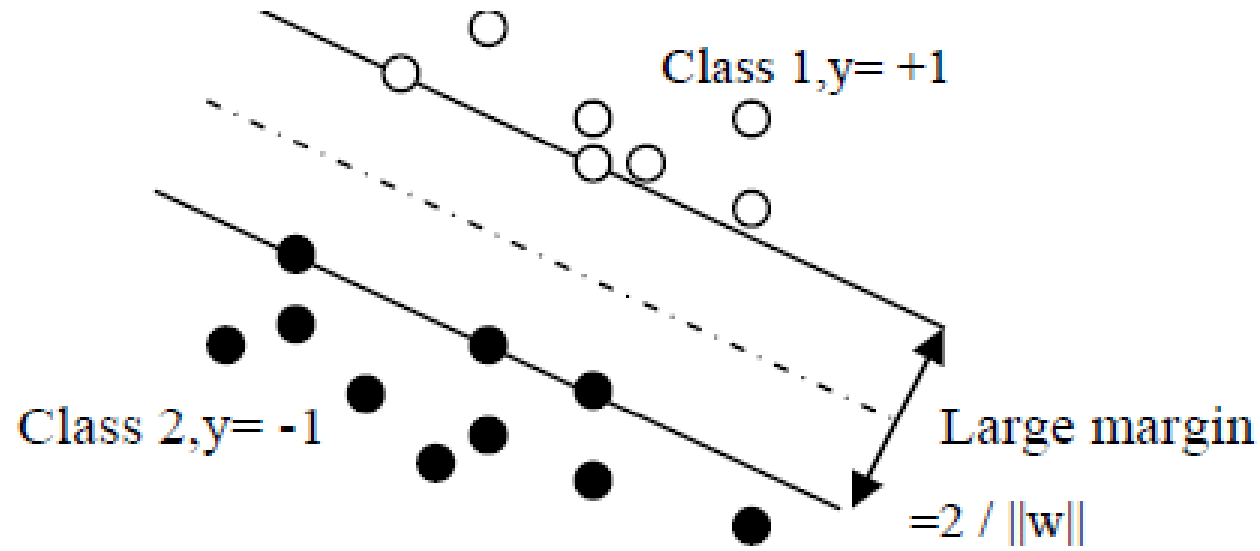
Timeline



Mathe – Michelle 😊😊😊

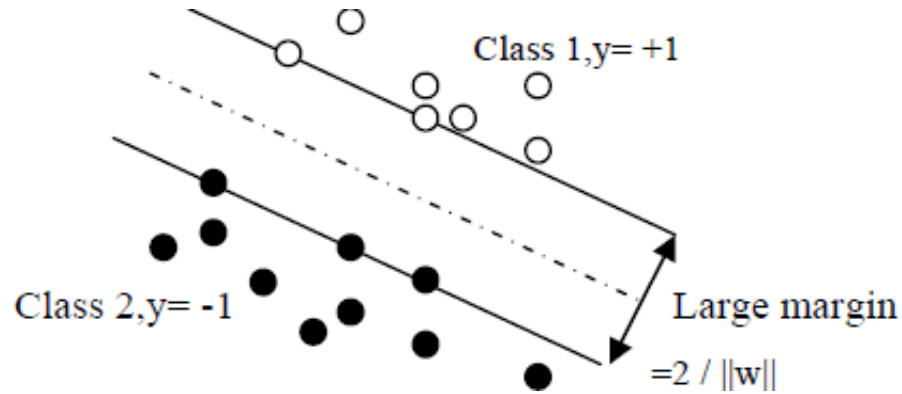


Support Vector machine



- based on statistical learning theory
- uses a decision function that defines a hyperplane
- hyperplane is the classifier, that categorizes data into two groups
- Phase 1: training phase
- Phase 2: generalization phase

Support Vector machine



- based on statistical learning theory

- uses a decision function that defines a hyperplane

- hyperplane is the classifier, that categorizes data into two groups

- Phase 1: training phase

- Phase 2: generalization phase

Planned analysis steps

- Implement a support vector machine
- Using the Dice-score-function to evaluate the performance of our SVM

Characteristics of the second Milestone

- Support vector machine labels pixels with ,cell nucleus' or ,background'

The purpose

- Use trained model to automatically segment cell nuclei images
- Compare the segmented images to the ground truth
→ using Dice Score

Second Milestone

Implement Support Vector machine

First Milestone

Implement Dice score

Third Milestone

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Our data

Timeline



Pre-processing

Original Image (Left) Vs. Gaussian Filtered Image (Right)



- Pre-processing steps:

1. Noise reduction: reduce the noise while preserving the edges
2. Super-pixel segmentation: join similar pixels together into a super-pixel region with a similar intensity value

- Desired effect:

1. Average local pixel intensity values to **reduce missing information** from original picture
2. Separate nuclei which appear fused for an improved segmentation

Pre-processing



Pre-processing steps:

1. Noise reduction: reduce the noise
2. Super-pixel segmentation: join pixels into a super-pixel region

Desired effect:

1. Reduce missing information from picture
2. Separate nuclei which appear fused

Planned analysis steps

Methods:

1. Gaussian filter
2. Gradient-ascend-based super pixel algorithms, e.g. Watershed

Characteristics of the third Milestone

- Improve Dice Score of segmentation method through better image quality

The purpose

- Achieve better registration results

Second Milestone

Implement Support Vector machine

First Milestone

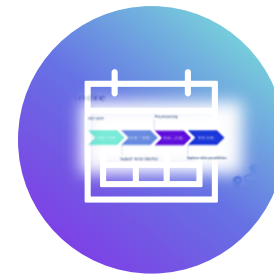
Implement Dice score

Third Milestone

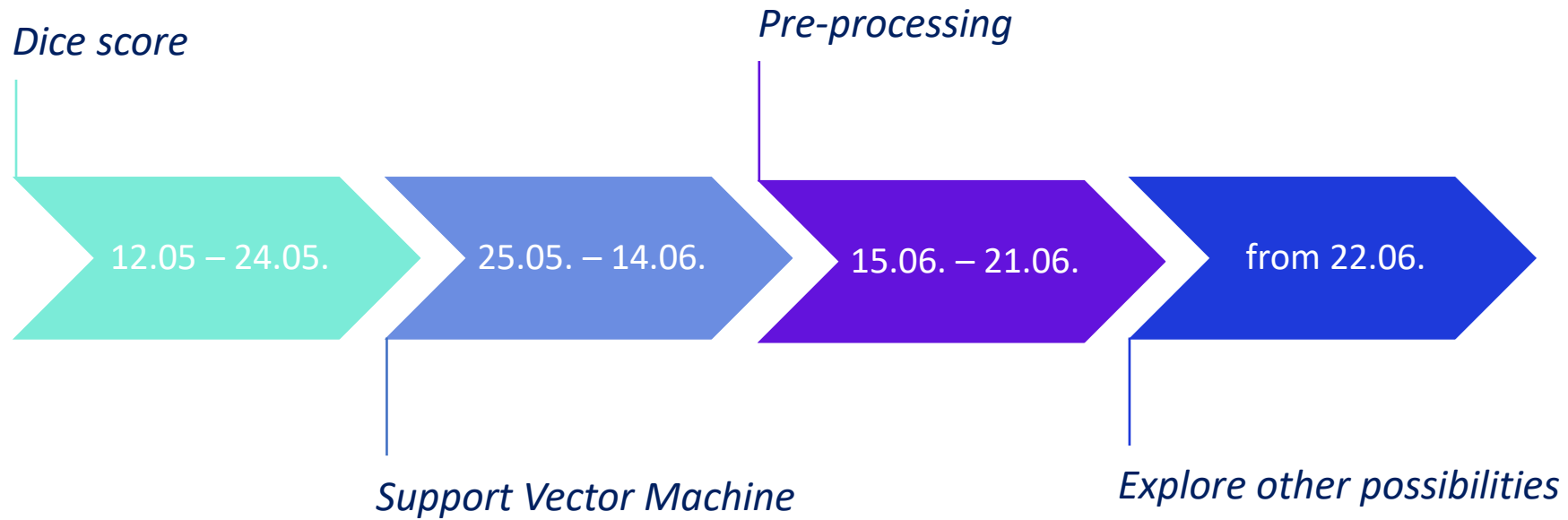
Implement Pre-processing

Our data

Timeline



Timeline



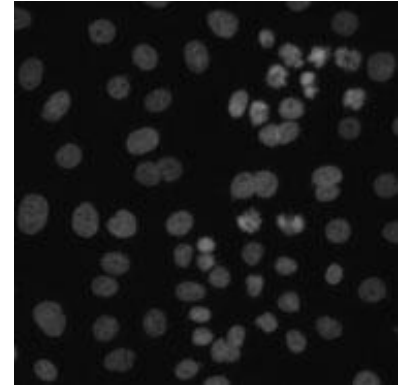
Other options to be explored

- testing our algorithm **on other data**, on which other algorithms were already tested on
→ compare results

e.g. Broad Bioimage Benchmark Collection 001 which was already used in

Nosova SA, Turlapov VE (2019) Detection of Brain Cells in Optical Microscopy Based on Textural Features with Machine Learning Methods. Program Comput Soft 45, 171–179

- try out more advanced pre-processing
e.g. anisotropic filtering



**Thank you for
listening!**

An abstract graphic on the right side of the slide, composed of several overlapping, rounded shapes in various shades of blue and purple. The colors range from a light sky blue to a deep, vibrant purple. The shapes are fluid and organic, creating a modern, artistic background element.