

AI use cases 2024



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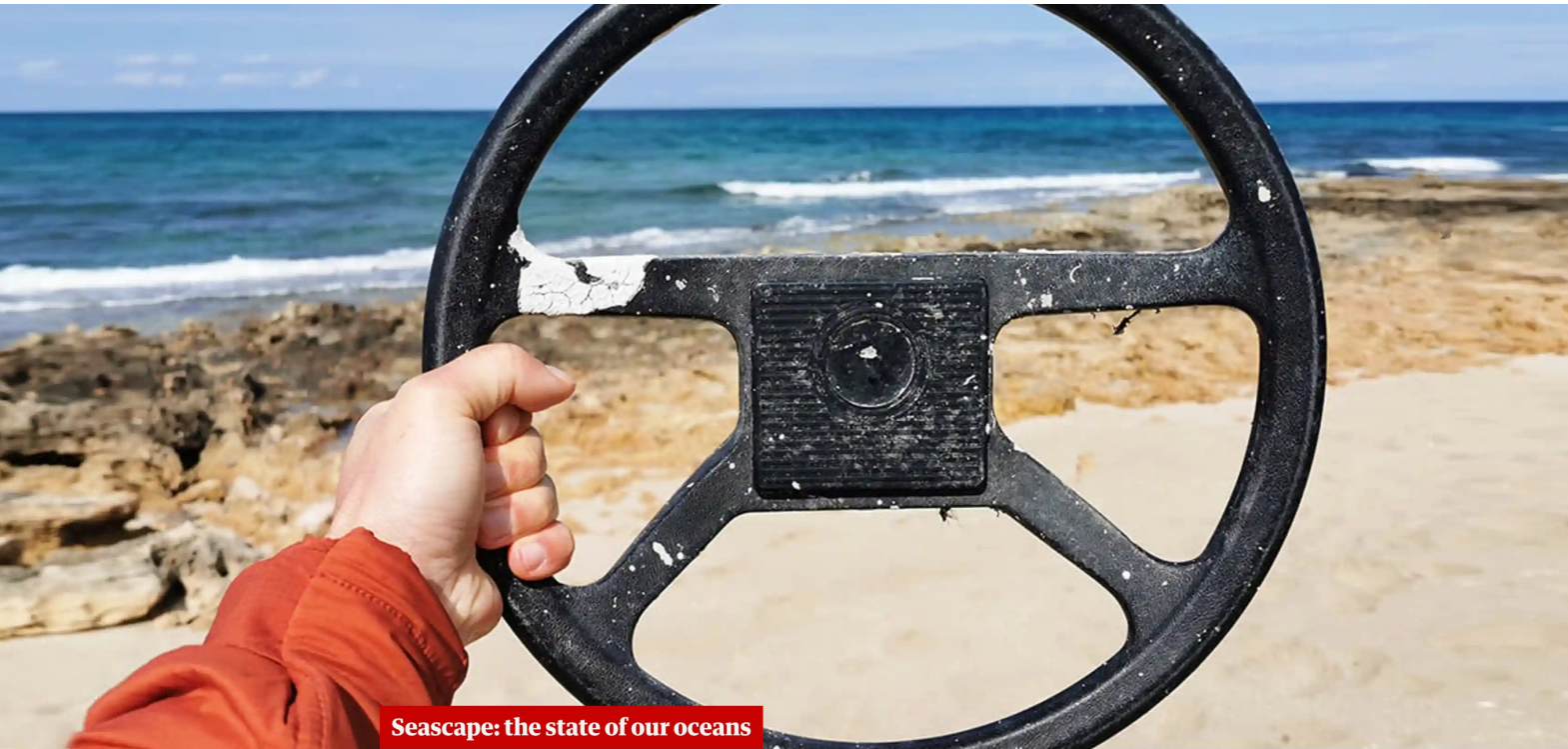




Case study 1: **THE OCEAN
CLEANUP**

**“Using AI to monitor
plastic density in the
ocean”**

“Plastic soup”: Every year, 14 million tons of plastic in the ocean



Seascape: the state of our oceans

Plastic never dies: the museum of vintage waste - in pictures

▲ A retro steering wheel found on the beach is among the items on display. Photograph: Archeoplastica



Archeoplastica



the next best thing to fresh!

St Ivel Fivepints is no ordinary milk powder. It tastes so good you can drink it on its own. You can use it everywhere you use fresh milk. For cooking, in tea, coffee and bedtime drinks. However, taste isn't the only advantage Fivepints has. Take a look at the pack. It's calibrated so you can see exactly how much you need to make each pint.

It has a resealable lid, so months and months. And it comes in an unb lightweight bottle which mak for the great outdoors. In fact, with St Ivel Five need never run out of milk as To make a pint: Add a ½ pint of warm water, whisk, with cold. Buy some the next time shopping.

SURF

<https://www.theguardian.com/>

<https://www.archeoplastica.it/>

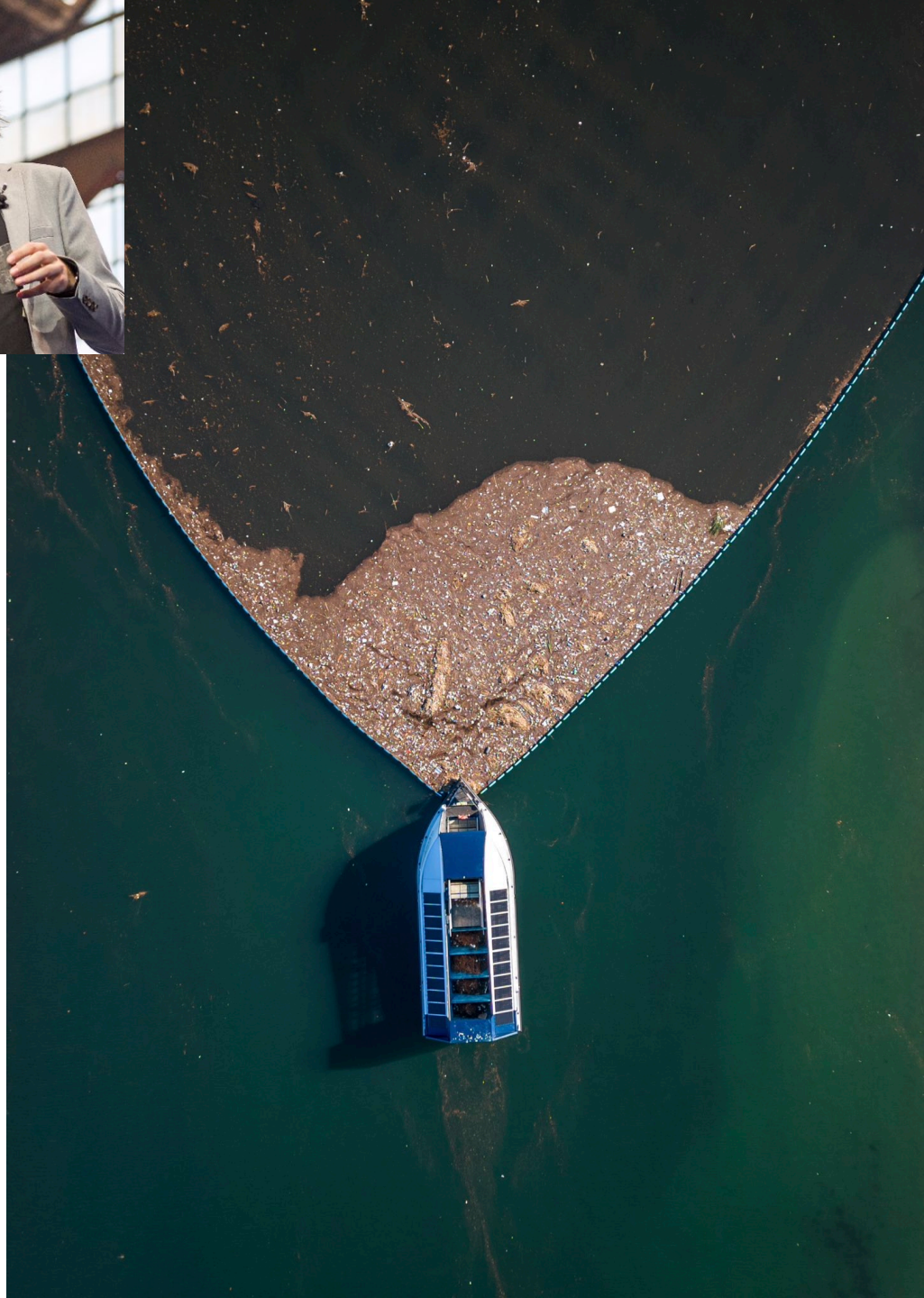
About The Ocean Cleanup



It all began when 16-year-old Boyan Slat (now CEO) was diving in Greece and saw more plastic bags than fish.

The Ocean Cleanup was founded to remove the plastic soup from the oceans.

How AI can come into play?



Idea: Using AI to monitor plastic density in the ocean

Goal: Have insights to **determine optimal locations** for operations

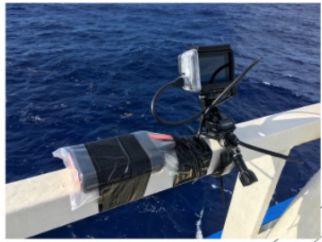
Potential impact: Knowing where the highest concentrations of plastic are located allows to **optimize resources** and maximize impact.

Conventional Method: Airplanes and trawls: labor intensive, **costly, and complex**

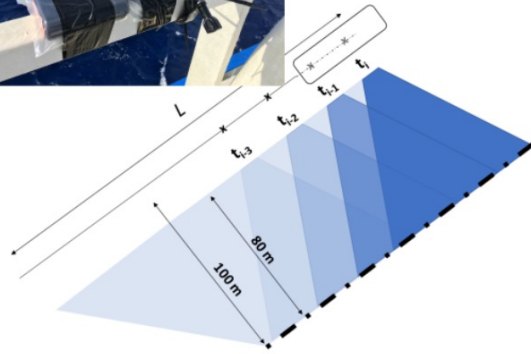
Idea: AI-based object detection software

Data collection & preparation

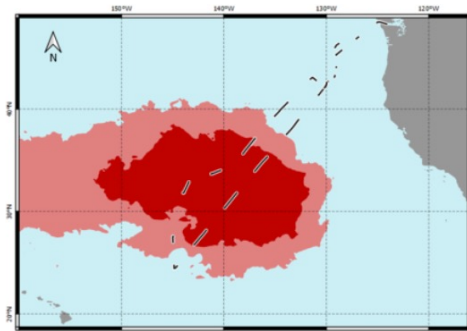
Data Collection:



GoPro
Photos with
GPS tag



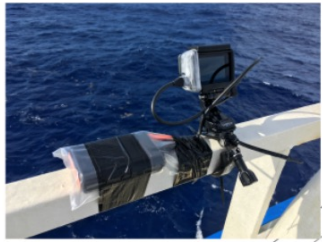
Over 100,000 photos collected



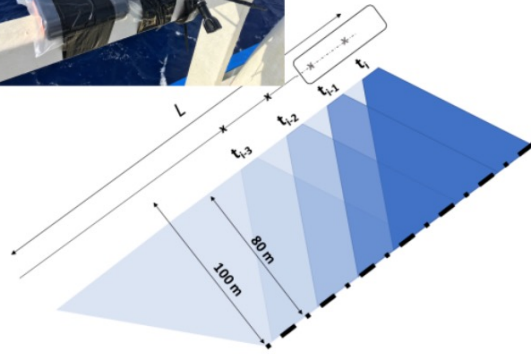
SURF

Data collection & preparation

Data Collection:

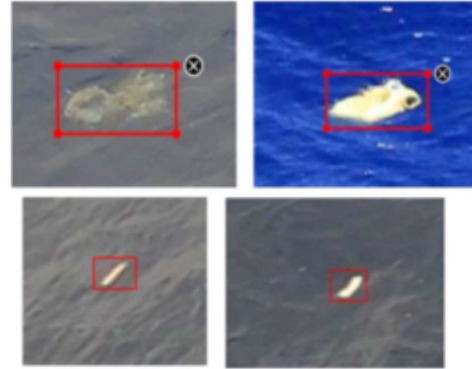


GoPro
Photos with
GPS tag

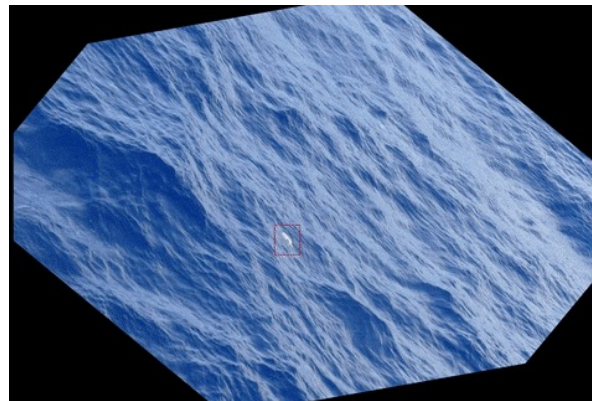


Data Preparation:

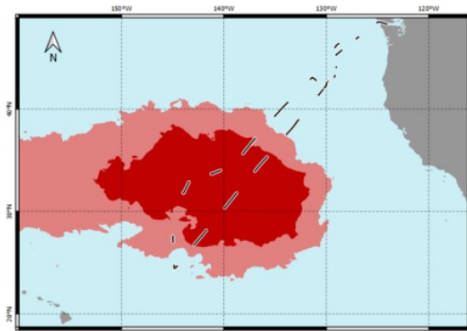
Labelling



Augmentation



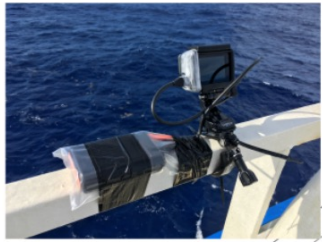
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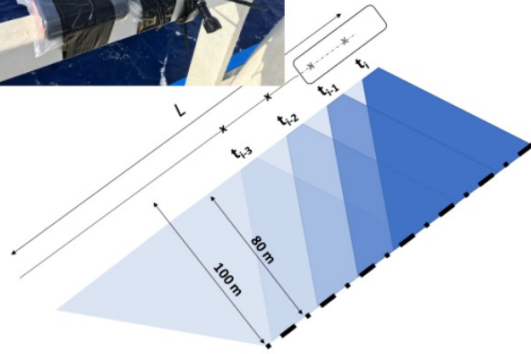
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Data collection & preparation

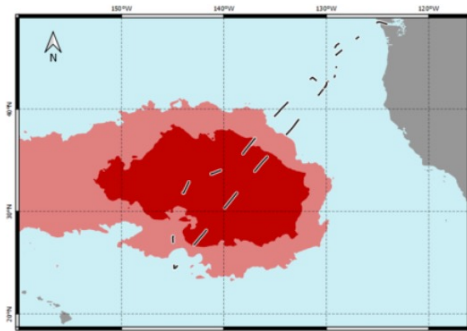
Data Collection:



GoPro
Photos with
GPS tag



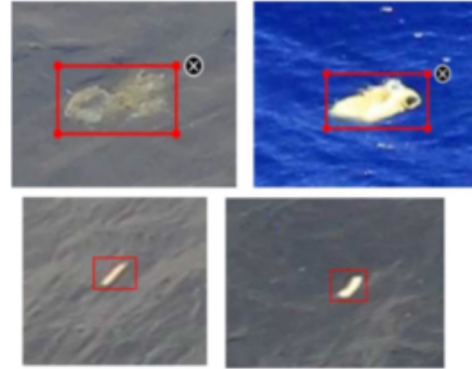
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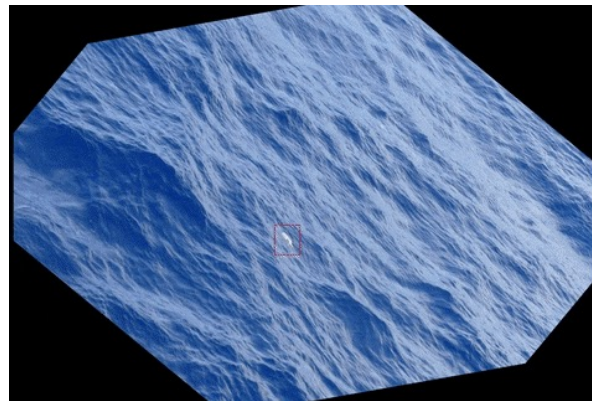
SURF

Data Preparation:

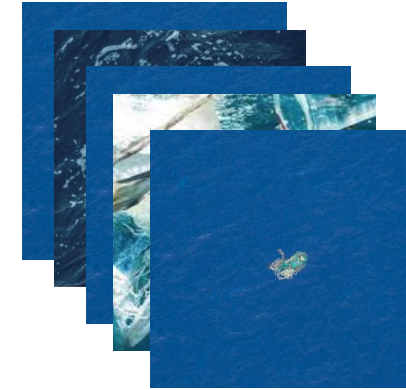
Labelling



Augmentation



Create Datasets:



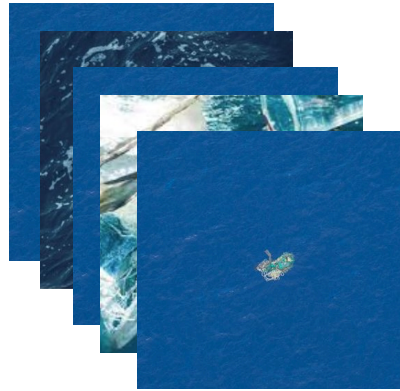
Training set:
18,589 images



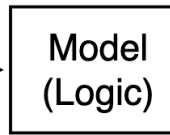
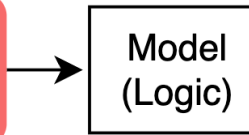
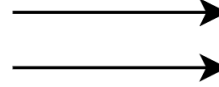
Test set:
739 images

Total of 600 GB of images

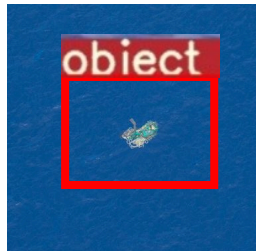
Training the AI algorithm



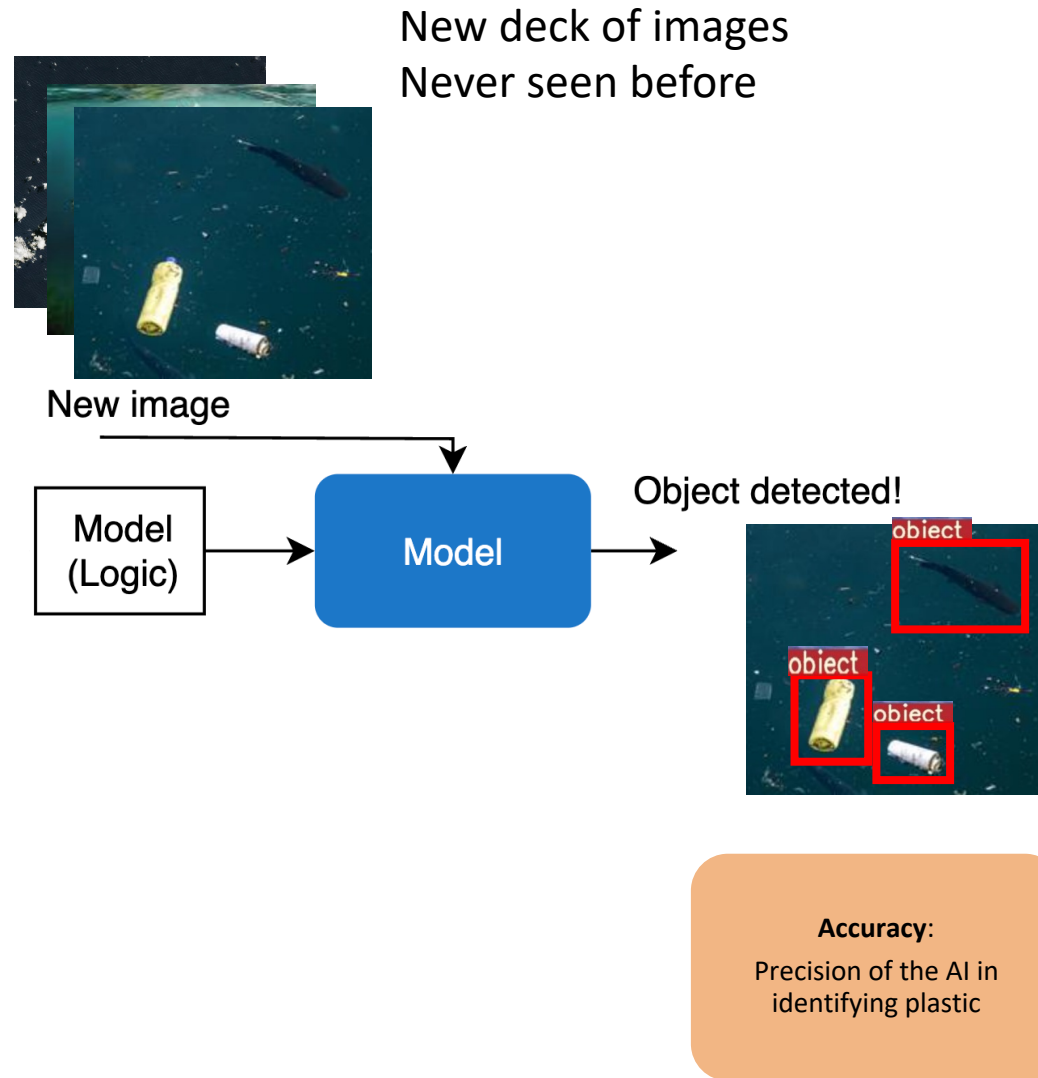
Training images



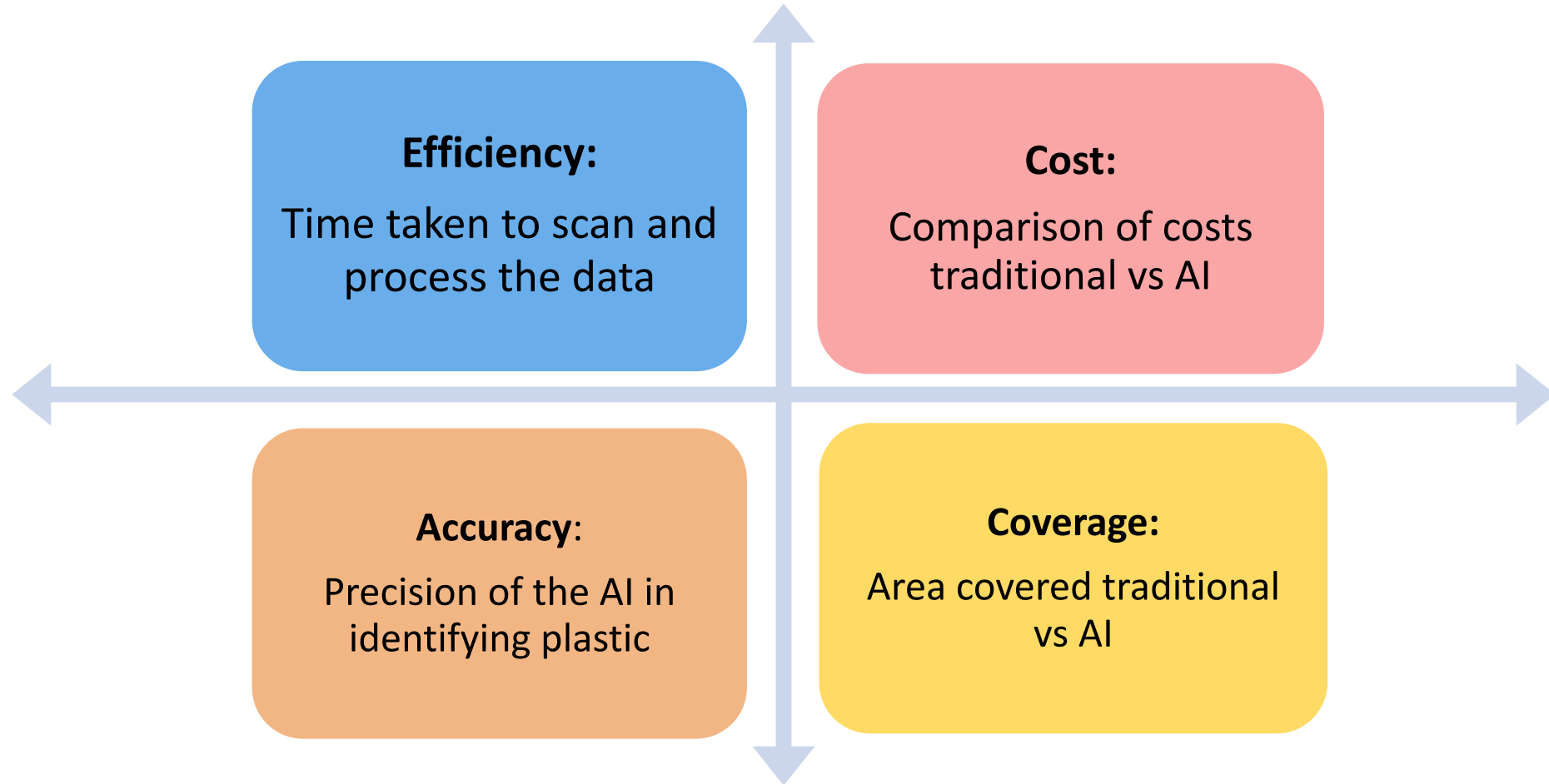
Detected object



Evaluating the AI algorithm



Metrics to evaluate the AI vs Traditional approach?



From Prototype to Production (& Monitoring): What are the challenges?

Training data “created” in a controlled environment

The model is trained and tested in HPC clusters

The prototype is handled by data scientist and engineers

Prototype is meant for a proof-of-concept, with iterative refinements based on feedbacks

Expect real-world data to be more messy (think of different light conditions)

Real-time object detection, integration with small devices and/or cloud

For a production-ready product, a user-friendly interface is essential

Regular checks for anomalies. Hw might become outdated, model becomes obsolete

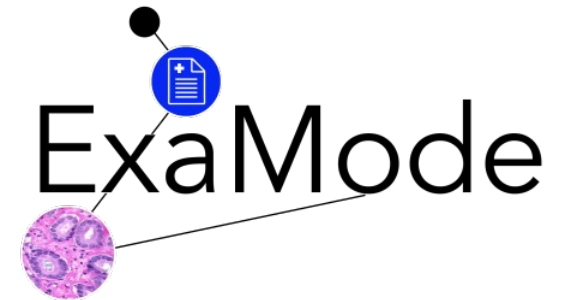


Case study 2:

**“Pre-trained LLMs for
caption generation of
histopathology images”**

Radboudumc
university medical center

AI advisor:
Bryan Cardenas Guevara
bryan.cardenasguevara@surf.nl



Digital pathology for cancer diagnosis:



Data collection

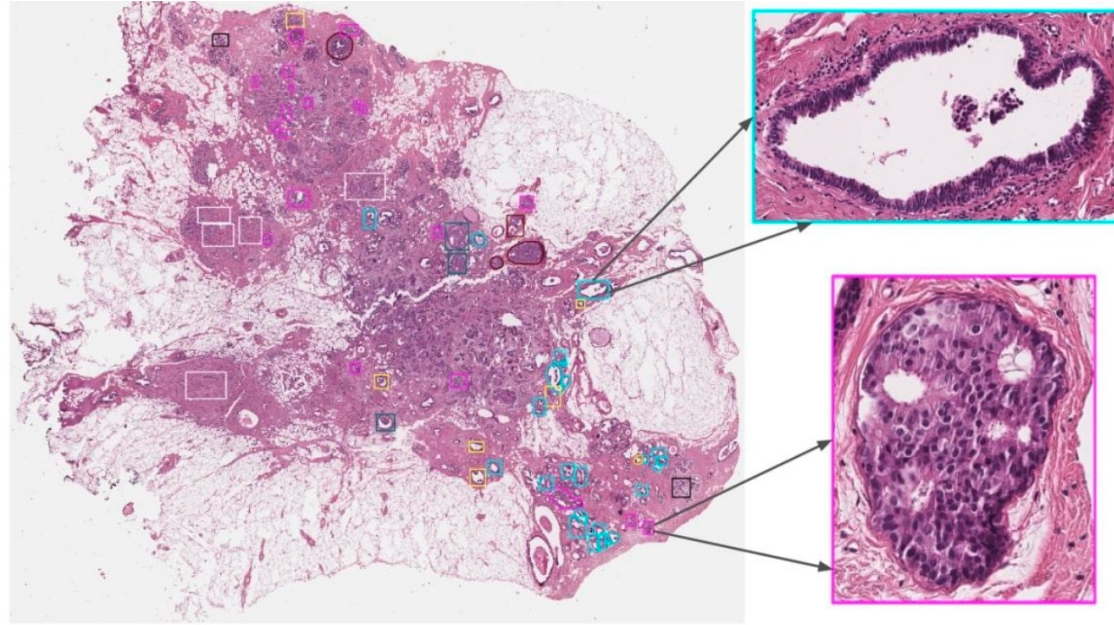
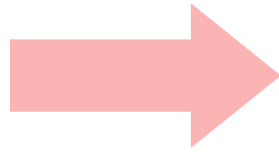


Image analysis
And
diagnosis

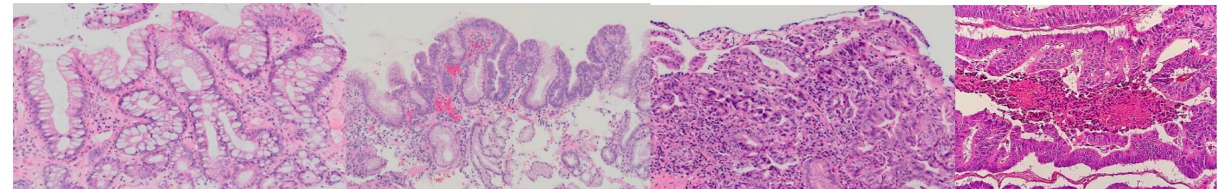
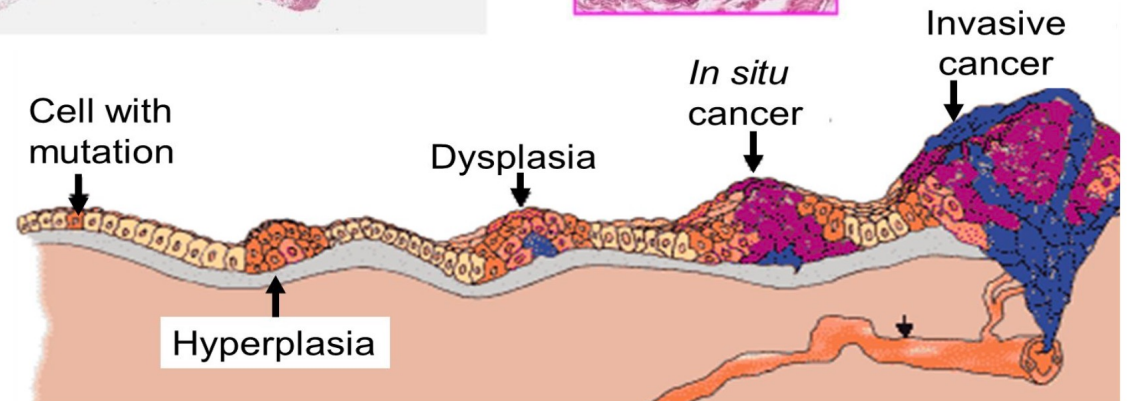


Image analysis: how ML is current used?

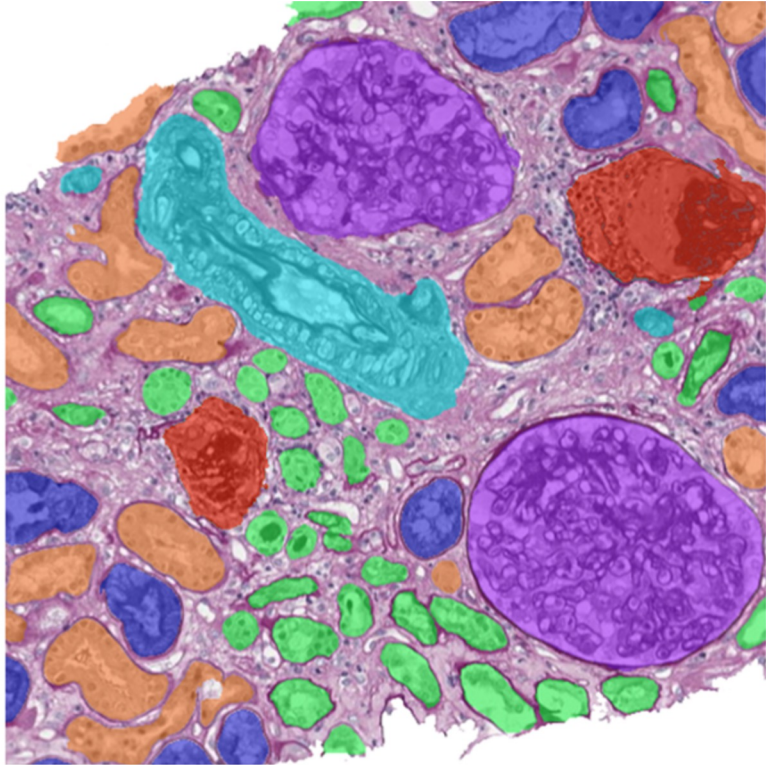
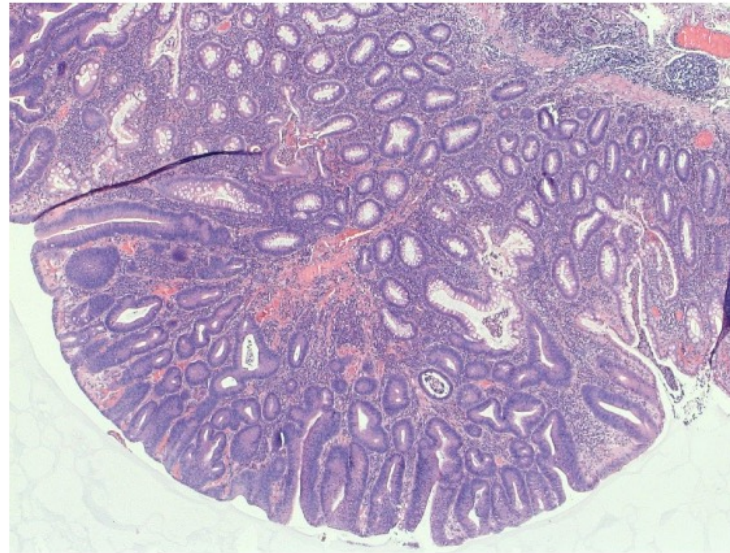


Image segmentation and multi-class classification



Normal

Hyperplasia

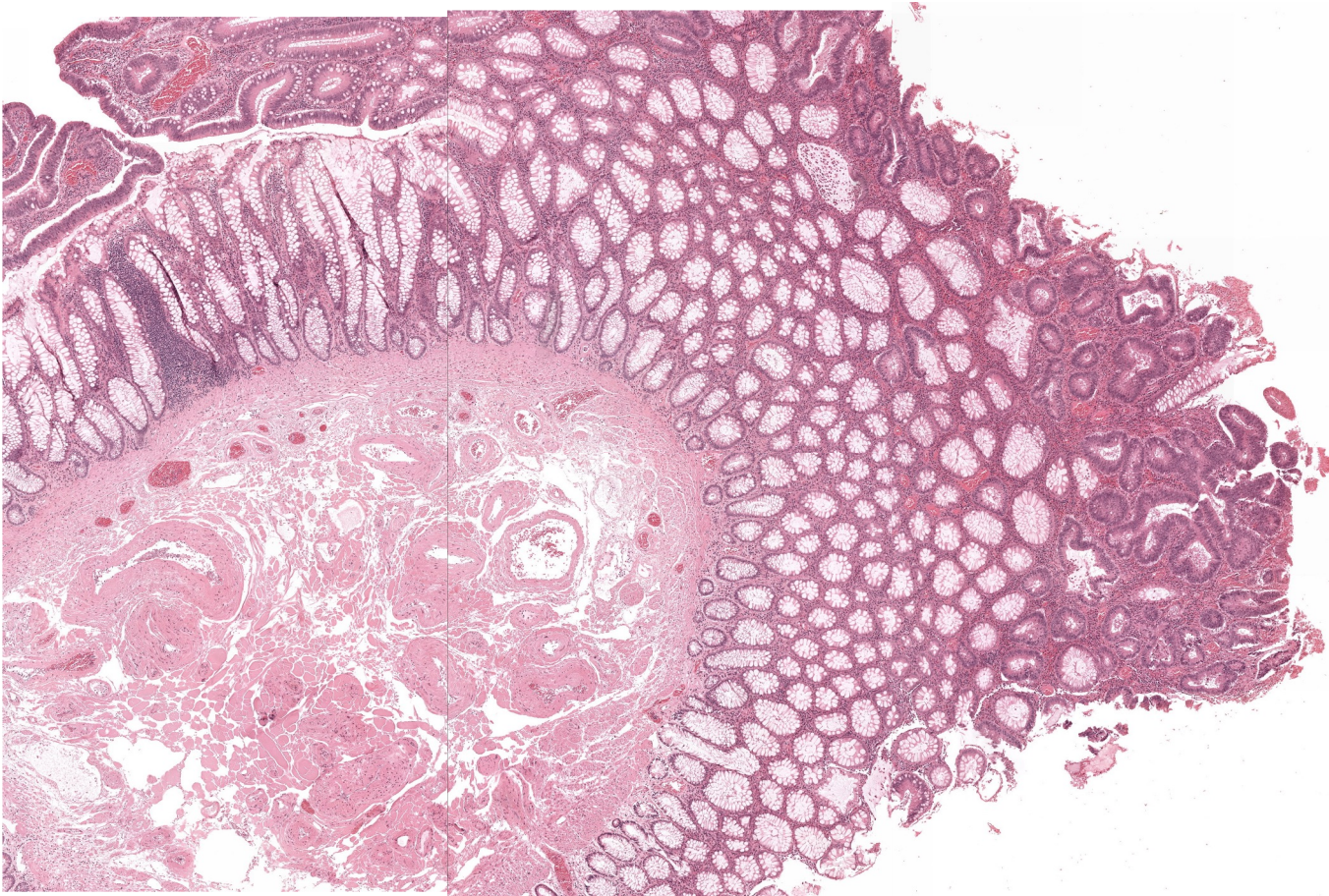
Low Grade Dysplasia

High Grade Dysplasia

Adenocarcinoma

Can we improve the text output to further help doctors in the **diagnosis**?

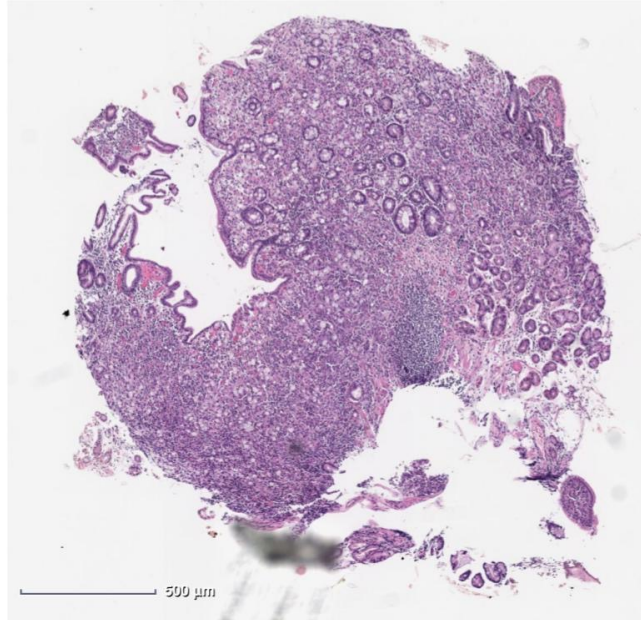
Histopathology data: Images and Text



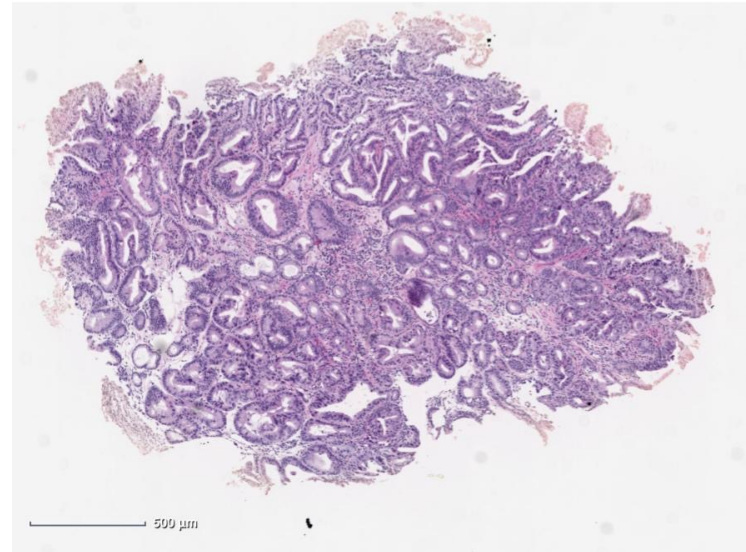
- ❖ “Colonbiopt 21-25 cm en 15 cm: in beide subrieken adenocarcinoma metastase, morfologisch en immuunhistochemisch geschikt voor het al bekend primaire ovariumcarcinoom van de patiënt.”
- ❖ 2 - adenoma tubulare con displasia bassa, focalmente, alto grado (moderato, focalmente grave)

Data collection: pair of whole slide images + captions

Pair <WSI-Caption> collected from two labs from two countries



A tumor tissue consisting of signet ring cell carcinoma cells is observed in the lamina propria. Signet ring cell carcinoma.



Tumor tissue consisting of medium-sized and irregular ducts is observed infiltrating in the superficial epithelium. Tumor cells are highly columnar, with nuclei aligned basolaterally and polarized. Well differentiated tubular adenocarcinoma.

- 5729 WSI-caption pair for training
- A subset of 569 WSI-caption for testing

Data preparation of the captions

Captions are:

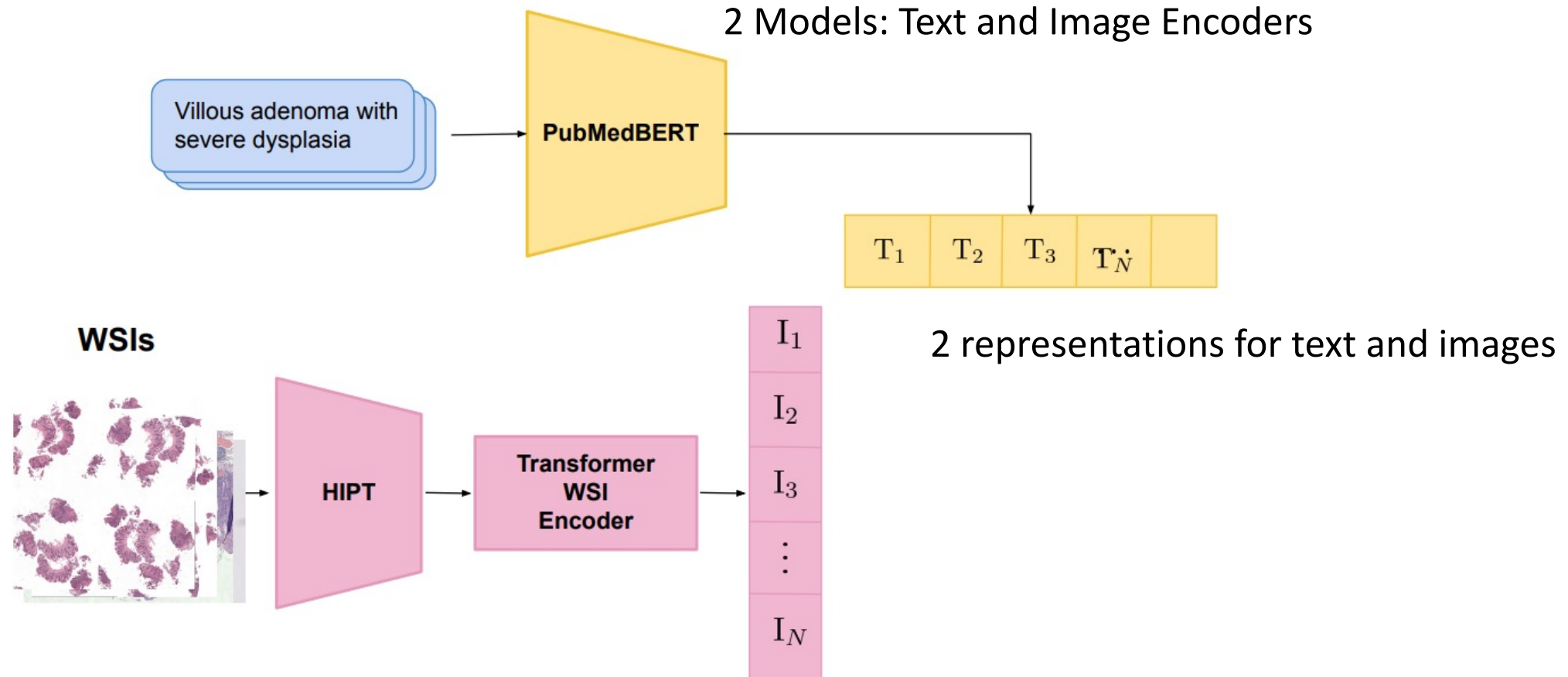
- Machine translated from two languages to English
- Preprocessed with GPT-3.5-turbo to remove irrelevant or noisy information
- ChatGPT was prompted with few examples of how to restructure the captions

Original Caption	GPT-3.5 Cleaned	Generated Caption
biopsies distal colon: chronic inflammation, in partially active and slightly histiocytary. no specific characteristics. the microscopic preparations from elsewhere have been requested for revision.	chronic inflammation, no specific characteristics.	no abnormalities, no dysplasia or malignancy. cyclic inflammation.
biopt colon transversum: adenocarcinoma.	adenocarcinoma.	Metastasis of adenocarcinoma best suited to primary process.
1) fragments of tubular adenoma with high degree dysplasia.	tubular adenoma with high degree dysplasia.	adenocarcinoma on villous adenoma. no lymphovascular invasion is identified. enced enced ED ED ED ED (failed caption generation)



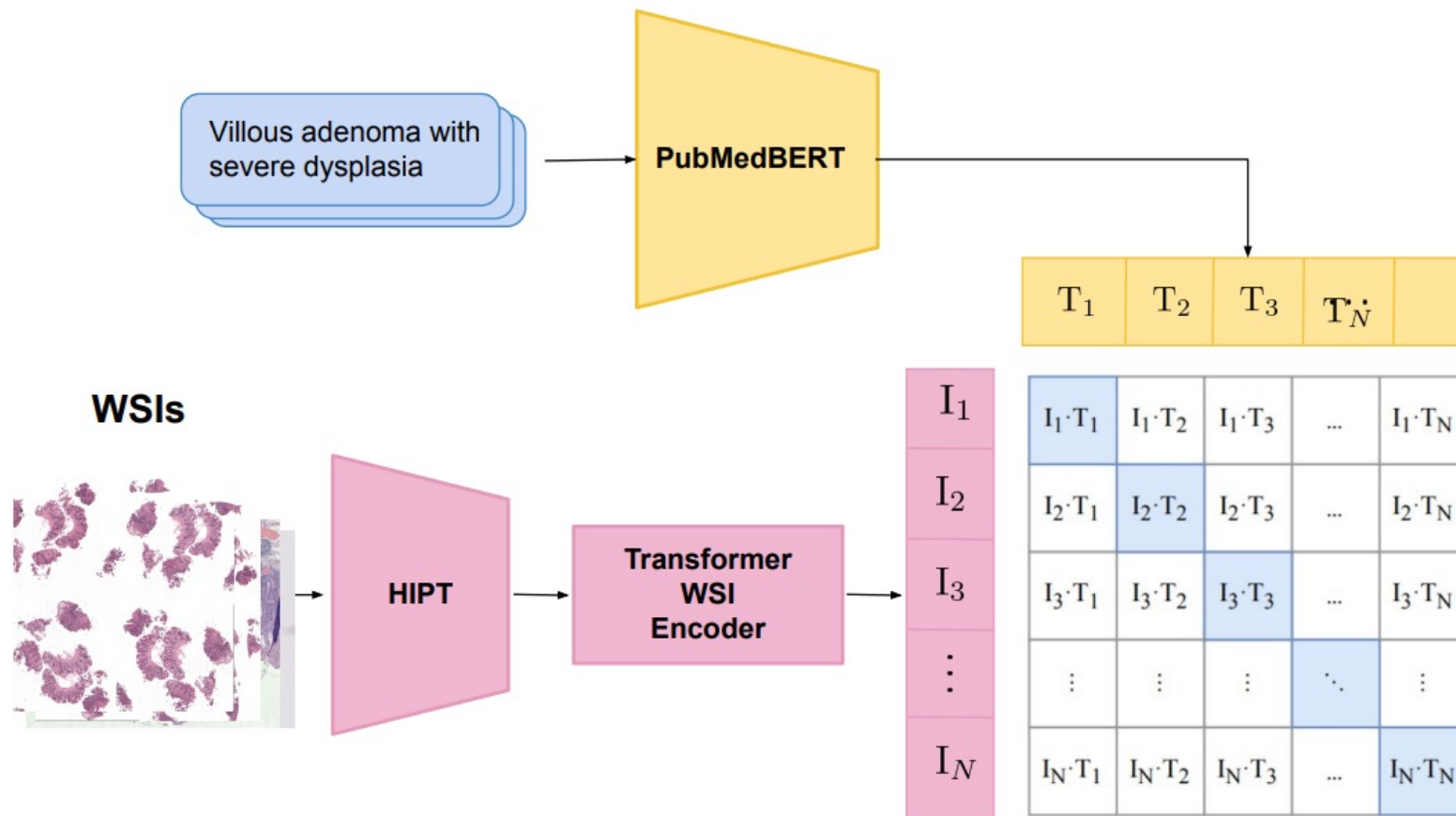
Model development: Pre-training phase

step 1: Encode information to numerical representation (vectors)



Model development: Pre-training phase

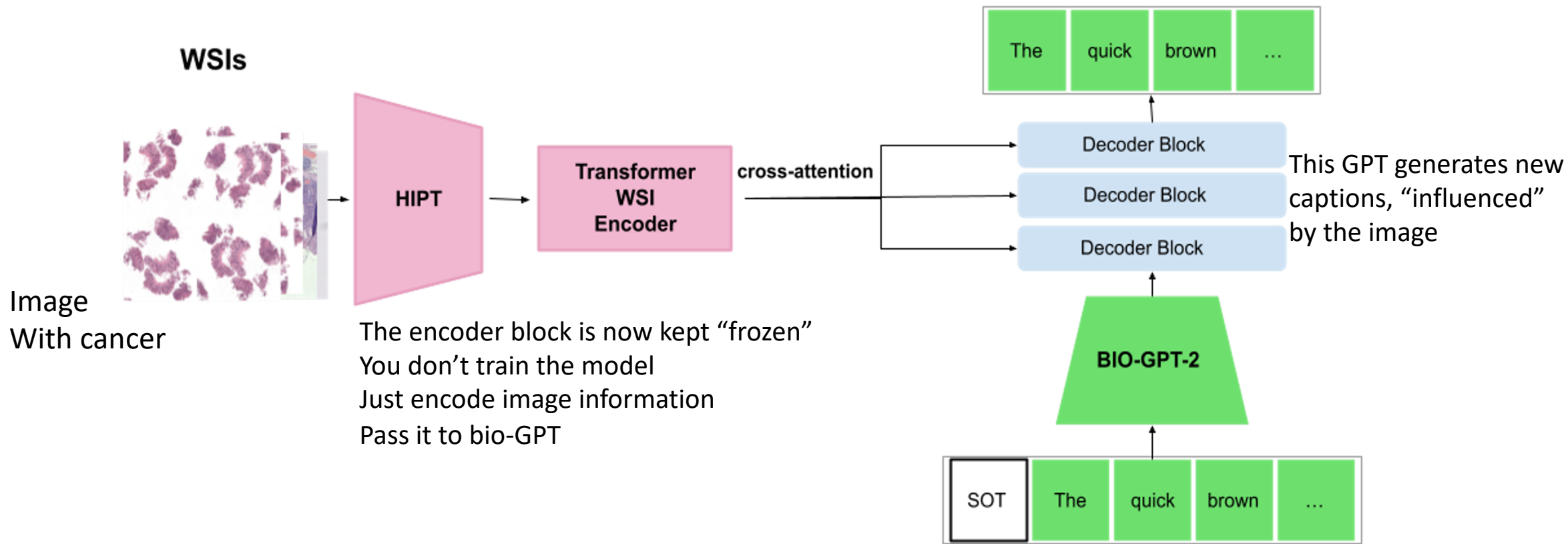
step 2: Train the model to maximize similarity



Vectors generated such that corresponding image-text receive a label 1 where non-related image-text receive a 0

Model development: Generation phase

"this image contains lesion of the form: cancer, high grade"



"this image contains lesion of the form:"

Evaluating the captions generated

Real: "polyep coecum bottom : tubular adenoma with locally high - grade dysplasia.",

Generated: "[CLS] polyep rectosigmoid : tubular adenomas with low grade (moderate) dysplasia and shows shows minor pseudomel some serrated high grade dysplasia.

Real: Polipe rectum : moderately differentiated adenocarcinoma.

Generated: rectum bioptes: localization of moderately differentiated adenocarcinoma.

Some references and resources:

- SURF <https://www.surf.nl/>
- EuroCC Netherlands <https://eurocc-netherlands.nl/nl/home-netherlands/>

The Ocean Cleanup:

- <https://www.surf.nl/en/case-computer-models-in-the-fight-against-plastic-soup>
- <https://theoceancleanup.com/updates/using-artificial-intelligence-to-monitor-plastic-density-in-the-ocean/>

Histopathology:

- <https://www.computationalpathologygroup.eu/projects/examode/>
- https://github.com/sara-nl/WSI_Captioning





Thank you!

 SURF, Amsterdam Science Park

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