INTERNATIONAL CONFERENCE OF THE PERMANENT COMMITTEE ON CADASTRE (PCC)





Al at GUGiK parctical use cases

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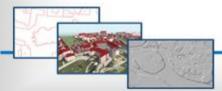


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AI IN PUBLIC ADMINISTRATION – KEY NUMBERS





92,5 %

digital public services make it easier to deal with official matters

51,1 %

Al will bring more benefits to society than harm

60,4 %

the state should use artificial intelligence to create digital public services

63,2 %

The state should more strongly regulate the use of artificial intelligence in business

88 %

citizens should have more control over how their data is used by public authorities

32 %

Al can speed up and automate administrative processes and reduce the time to deliver public services

Artificial Intelligence in public administration – The Polish Economic Institute, 2024

66,7 %

the state should invest more in digital public services, but only 27.2% are willing to finance these investments through tax increases

6,02 PLN (1, 40 EURO)

respondents are willing to pay per month for an application that would use AI to pre-fill forms required by a given office



AI IN PUBLIC ADMINISTRATION – CONCERNS AND HOPES





concerns about the use of AI:

- ✓ loss of privacy and threats to personal data protection
- √ lack of transparency of AI algorithms
- ✓ blurring of responsibility for decisions made



hopes related to the use of AI:

- ✓ improving public services and reducing waiting times
- ✓ improving accessibility of services for people with disabilities
- ✓ reducing Human Errors
- ✓ automation Repetitive Tasks and Processes
- ✓ Smoothly handles Big Data
- ✓ AI-Enabled Digital Assistants
- ✓ it is Full-Time Available



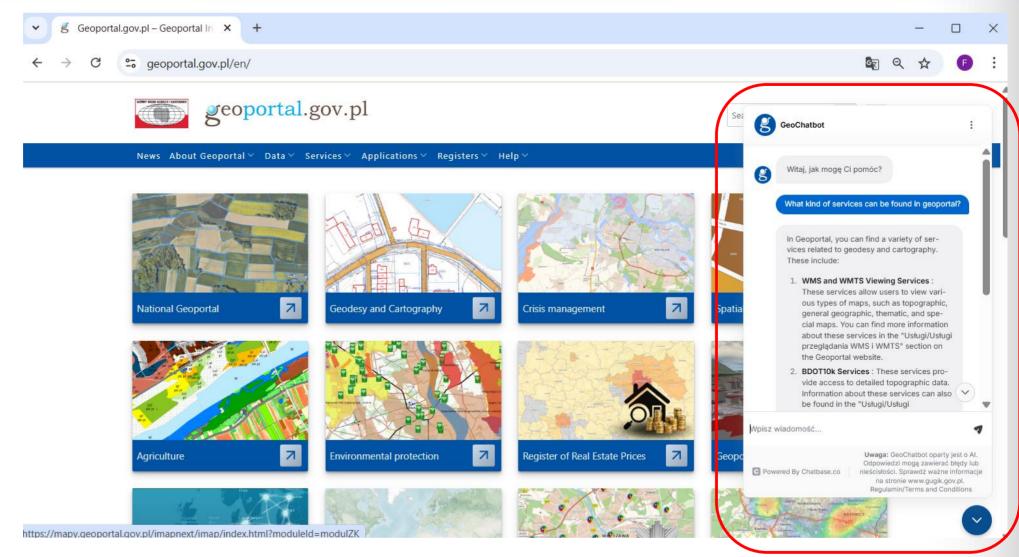
GeoChatbot in geoportal service





Digital Assistants based on AI technology

- ✓ AI-based chatbots reduce the need for unnecessary customer service staff.
- ✓ Helps in addressing routine and simple customer queries.
- ✓ Chatbot smartly handle customers and provide timely and accurate information on demand.



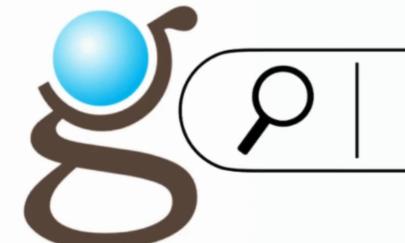
GeoChatbot in geoportal service

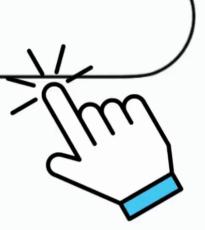






geoportal.gov.pl

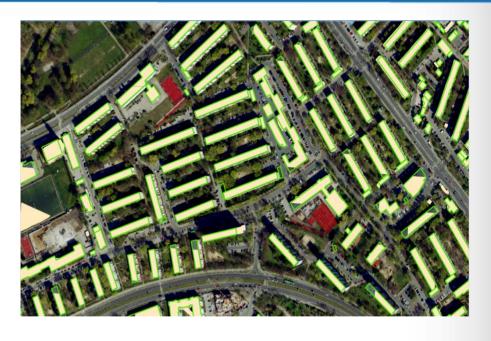












- ✓ Program: The INFOSTRATEG project is implemented by the National Center for Research and Development.
- ✓ Estimated budget: 40 mln PLN (~ 8,6 mln EUR)
- ✓ Project goal: creation of tools based on artificial intelligence algorithms enabling automatic detection of topographic objects using photogrammetric data.

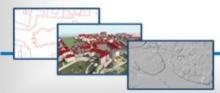




Expected effects:

- ✓ automation of the BDOT10k database update process
- √ reduction of public funds expenditure
- ✓ shortening of update time
- ✓ increasing the frequency of register updates
- ✓ improvement of the database quality (enabling constant access to reliable data for citizens, services and institutions using the database to carry out tasks related to, among others, crisis management, spatial planning, environmental analyses and investment processes









Project steps:

training data preparation

model training level

initial application

technical documentation

phase I

buildings, structures
and facilities
communication network



min. 85%

phase II

land development network land cover water network



min. 90%

—

min. 85%

phase III

complete training data for all BDOT10k classes



min. 95%

initial application enabling the launch of object detection for the indicated BDOT10k categories and generating reports with the appropriate statuses: "new object", "changed object", "identical object"

final version of the application

full technical documentation with instructions and directions for model and application development



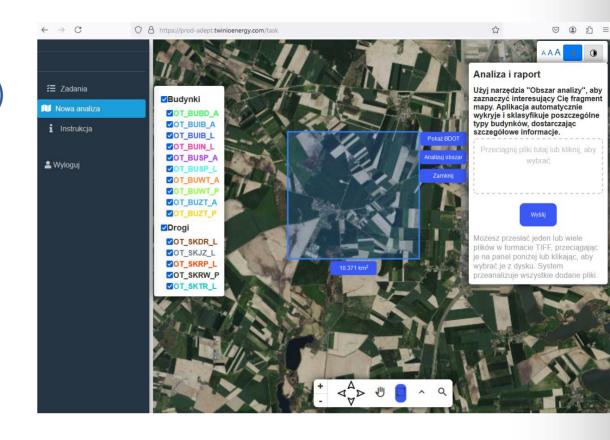




phase I

contractor 1: PROA TECHNOLOGY Sp. z o.o, LPE POLAND Sp. z o.o

- ✓ ADEPT application with graphical user interface (GUI)
- ✓ detection level:
 - above 80% 3 classes
 - 50% 70% 11 classes
 - 0% 1 class
 - not included in the application 9 classes
- ✓ detection time less than 3 min for an area of 5 km²





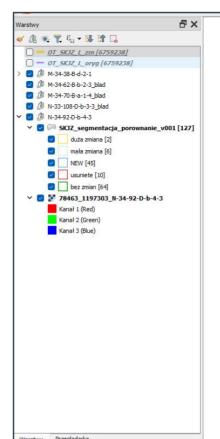




phase I

contractor 2: GISPRO TECHNOLOGIES Sp. z o.o., Maritime University of Technology in Szczecin

- ✓ GUI application in the form of 4 plugins for QGis
- ✓ detection level:
 - above 80% 3 classes
 - 40% 50% 18 classes
 - 0% 2 classes
 - not included in the application 1 class
- ✓ detection time less than 3 min for an area of 5 km²











phase I

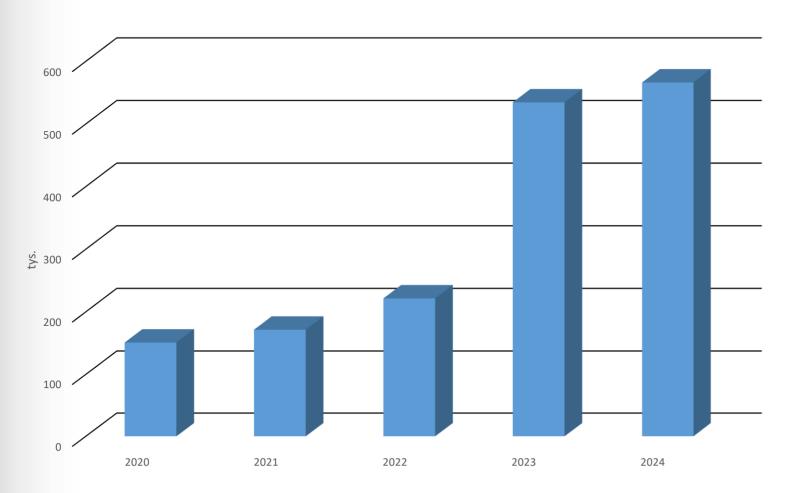
contractor 3: REASONFIELD LAB Sp. z o.o, University of Łódź

- ✓ linux-based GEOAI application
- ✓ detection level for areas smaller then 5 km2:
 - above 80% 2 classes
 - not included in the application 22 classes
- ✓ no possibility to specify the detection time
 (the application does not perform a comparison for an area of 5 km2)



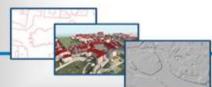






The number of aerial images taken is growing in an exponential rate due to:

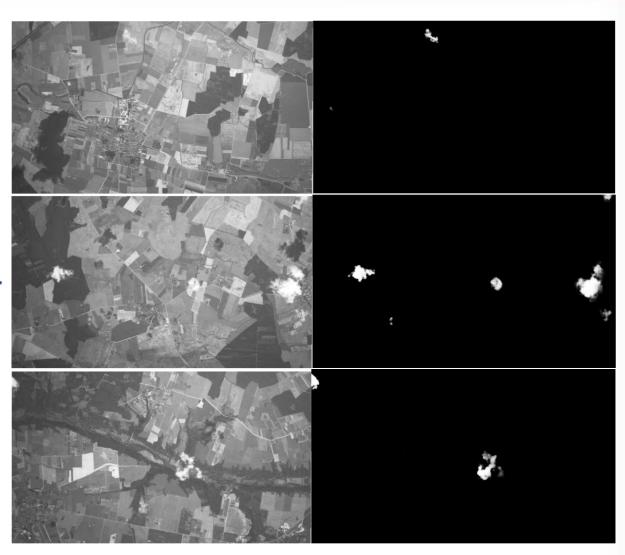
- ✓ the lower GSD,
- ✓ more frequent data acquisitions, new aerial platforms and sensors (oblique cameras)
- ✓ new products that requires higher overlaps (true-ortho, 3D meshes).







Examples of images with clouds (left) and corresponding probability images that predict the artificial neural network (right).

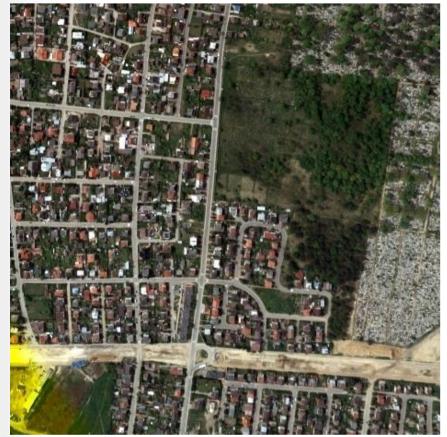








Examples of photos with discoloration













Example of a blurred image detected (left)











Example of reflectors, burn-in







Key technical challenges:

- ✓ Huge diversity of aerial images. The requirement for the solution to be effective for all photos taken, regardless of the pixel size, type of land cover, season, or the angle of the sun's rays.
- ✓ Errors occurring in the photos also vary in appearances. For example, clouds can have any shape, size, and thickness, which means that the image of the Earth's surface may be completely obscured or partially visible through the cloud.
- ✓ Lack of large-scale training data, more than 98% of images delivered are without errors.

The basic condition for the development of an effective model of radiometric QA in aerial photos based on machine learning techniques is the **preparation of a sufficiently large and varied set of training data**. The training was done on more than 10 thousand aerial images.







✓ Blur

Deep learning model uses full resolution of the image to detect even small blur

✓ Radiometric errors

Detection of discoloration, burn-in and glare, reflectors

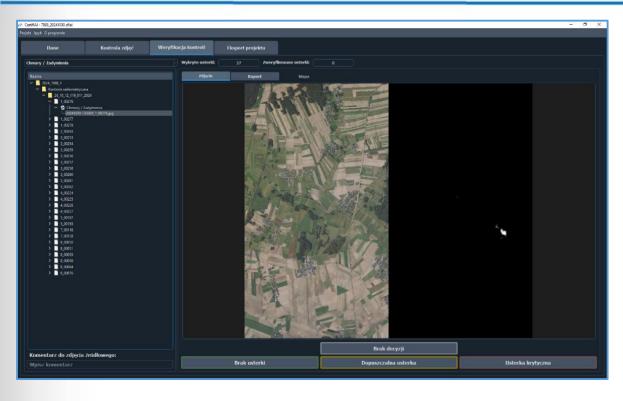
✓ Clouds, clouds shadows and smoke
Important feature can be obstructed by such elements, so it is important to correctly asse their impact on the datasets

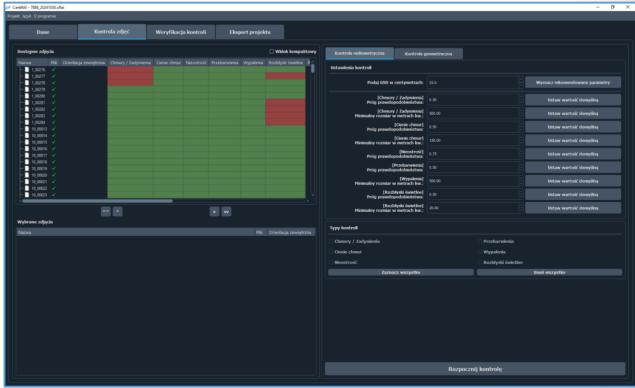
Kontrola radiometryczna Kontrola g	geometryczna			
Ustawienia kontroli				
Podaj GSD w centymetrach:	5.0		Wyznacz rekomendowane parametry	
[Chmury / Zadymienia] Próg prawdopodobieństwa:	0.50		Ustaw wartość domyślną	
[Chmury / Zadymienia] Minimalny rozmiar w metrach kw.:	50.00		Ustaw wartość domyślną	
[Cienie chmur] Próg prawdopodobieństwa:	0.50		Ustaw wartość domyślną	
[Cienie chmur] Minimalny rozmiar w metrach kw.:	150.00		Ustaw wartość domyślną	
[Nieostrość] Próg prawdopodobieństwa:	0.75		Ustaw wartość domyślną	
[Przebarwienia] Próg prawdopodobieństwa:	0.50		Ustaw wartość domyślną	
[Wypalenia] Minimalny rozmiar w metrach kw.:	100.00		Ustaw wartość domyślną	
[Rozbłyski świetlne] Próg prawdopodobieństwa:	0.50		Ustaw wartość domyślną	
[Rozbłyski świetlne] Minimalny rozmiar w metrach kw.:	10.00		Ustaw wartość domyślną	
Typy kontroli				
☐ Chmury / Zadymienia		☐ Przebarwienia		
☐ Cienie chmur		☐ Wypalenia		
□ Nieostrość Zaznacz wszystko		Rozbłyski świetlne	Usuń wszystko	
Zaznacz wszystko			osun wszystko	
Rozpocznij kontrolę				





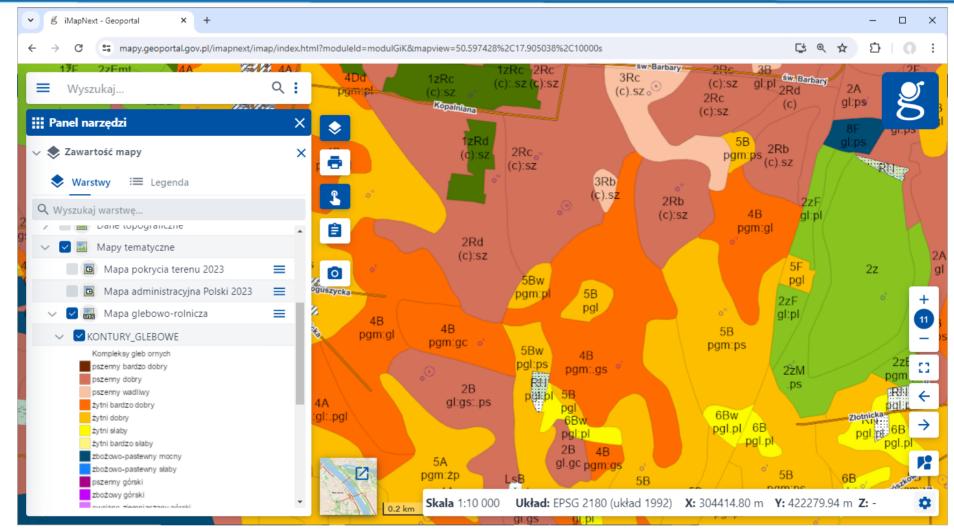












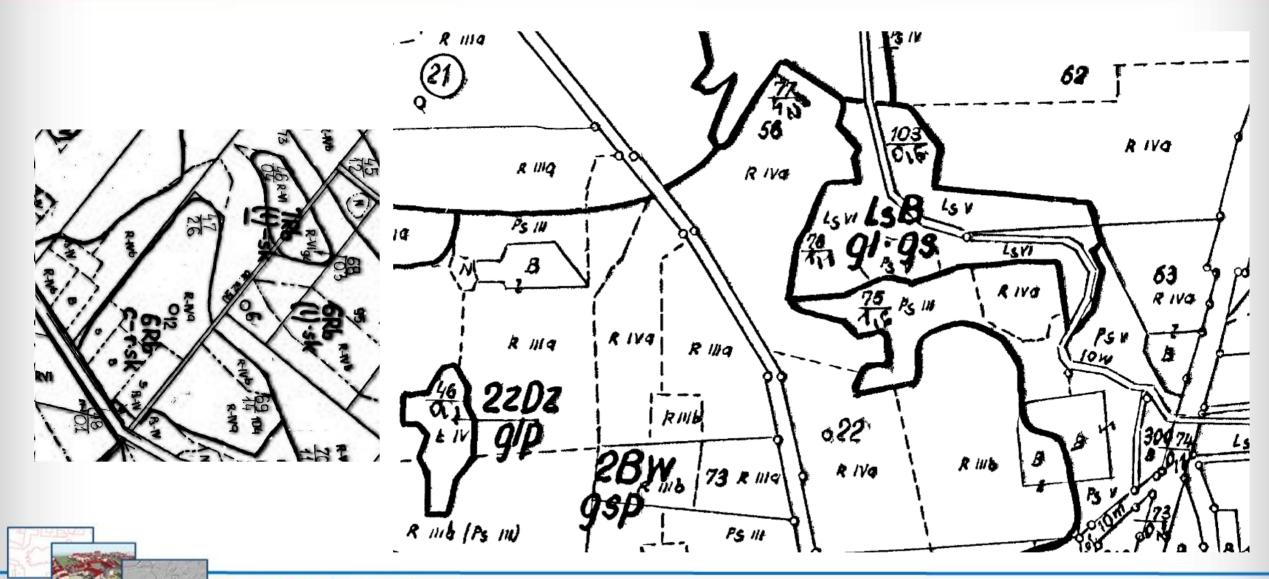
https://mapy.geoportal.gov.pl/wss/service/pub/guest/MapaGlebowoRolnicza/MapServer/WMSServer

WMS

WMTS



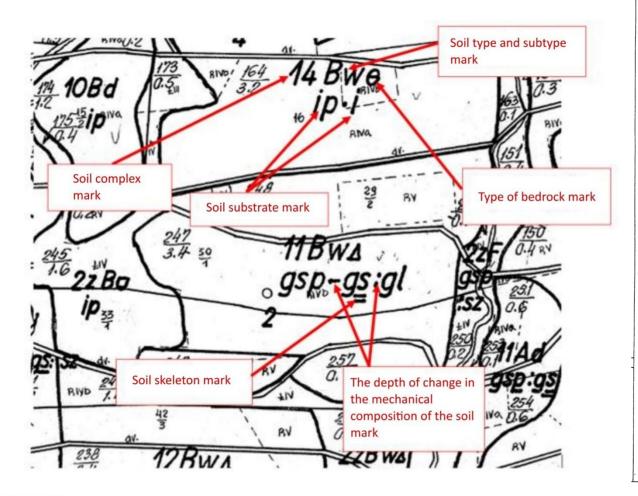








CONTENT OF THE SOIL MAP 1:5000



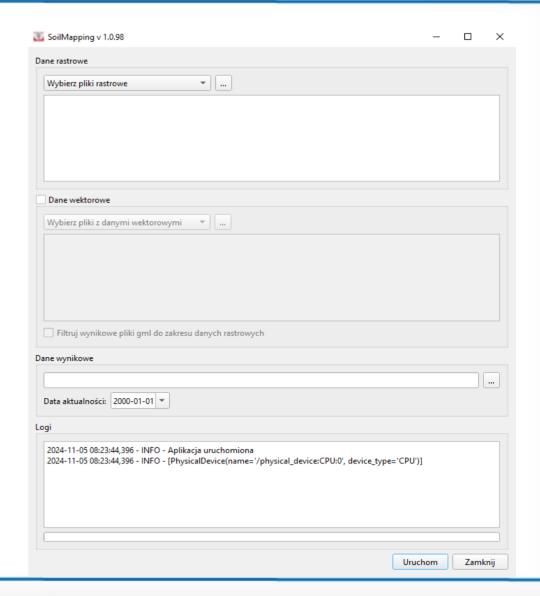
OZNAC	ZENIA		
KOMPLEKSY	RODZAJE I GATUNKI		
1 – kompleks pszenny bardzo dobry	Ź – żwiry piaszczyste		
2 – kompleks pszenny dobry	ŻG – źwiry gliniaste		
3 – kompleks pszenny wadliwy	pl – piaski lužne		
4 – kompleks żytni bardzo dobry	DS — piaski stabo gliniaste		
5 – kompleks żytni dobry	pgl – piaski gliniaste lekkie		
6 – kompleks żytni staby	PQM – piaski gliniaste mocne		
7 – kompleks żytni najstabszy	gl – gliny lekkie		
8 – kompleks zbożowo - pastewny mocny	QS - gliny średnie		
9 – kompleks zbożowo - pastewny staby	GC – gliny ciężkie		
14 gleby orne przeznaczone pod użytki zielone	ptZ – pyty zwykte (gleby pyłowe lekkie i śr.)		
1z – użytki zielone bardzo dobre i dobre	pti - pyty ilaste		
2z – użytki zielone średnie	ls – lessy i utwory lessowate gleby lessowe i lessowate lekkie i średnie		
3z – użytki zielone stabe i bardzo stabe RN – gleby rolniczo nieprzydatne nadające się pod zalesienie	[_ lessy ilaste-gleby lessowe i lessowate		
Iz - tereny zabudowane o zwartej zabud.	i – ity (gleby ilaste bardzo ciężkie)		
Ls – lasy	P – oznaczenie pylastości gleb		
W -wody			
WN –wody nieużytki	RODZAJE GLEB TORFOWYCH		
N – nieużytki rolnicze	I TORFOWO - MUŁOWYCH		
	N — torfy niskie		
TYPY	mt - gleby mutowo-torfowe		
	tm — gleby torfowo-mutowe		
A – gleby bielicowe i pseudobielicowe			
B – gleby brunatne właściwe	OZNACZENIA DODATKOWE		
Bw – gleby brunatne wyługowane i bru – natne kwasne			
D – czarne ziemie właściwe	• - podłoże zalega płytko (do 50 cm)		
Dz – czarne ziemie zdegradowane i gleby szare	- podłoże zalega średnio głęboko (50 - 100 cm)		
	- podłoże zalega głęboko (100 - 150 cm)		
G — gleby glejowe E — gleby mułowo-torfowe i torfowo	O 2 lokalizacja i numer odkrywki opisanej przez redoktora terenowego O 3 lokalizacja i numer odkrywki, z której pobrano próbki do analiz		
T - gleby torfowe i murszowo-torfowe	2 - numer konturu		
M _ gleby murszowo - mineralne	5,2 - powierzchnia		
i murszowate	≈ – gleba podlega zalewom rzecznym.		





The input data:

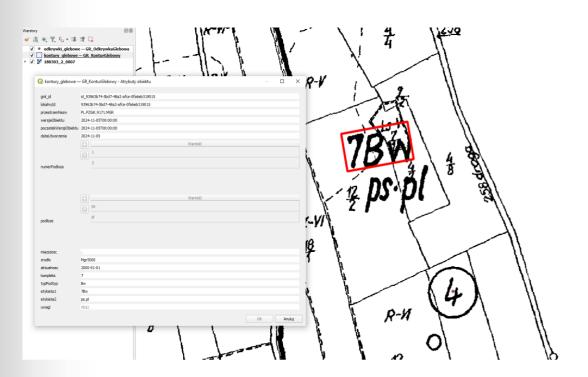
- ✓ georeferenced raster data in GeoTIFF format in the coordinate system EPSG:2180.
- ✓ vector data on soil contours and outcrops in GML, XML and SHP format (optional).



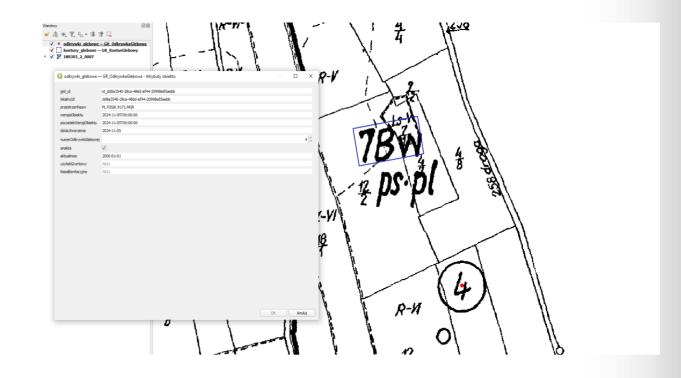




Soil contours:



Soil outcrops:







✓ The application uses the developed machine learning models trained on a very large volume of training data (**The Convolutional Neural Network (CNN)** architecture of the U-Net).

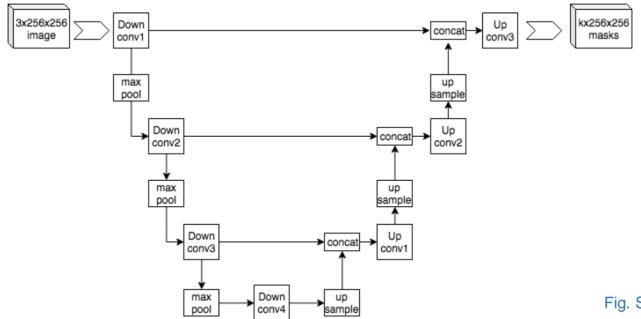


Fig. Source: https://en-m-wikipedia-org

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