kadaster

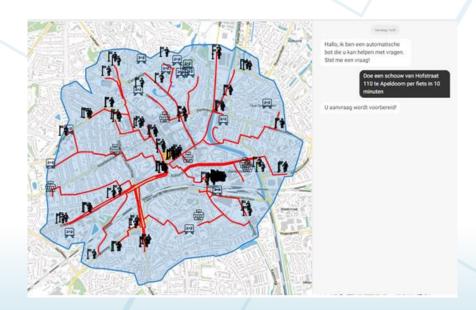
# How Al shapes land registration in the Netherlands

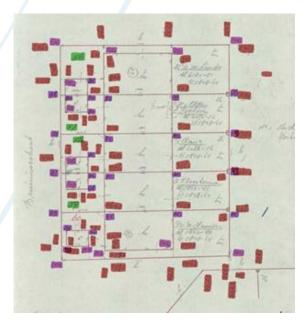
Wim Florijn



### Use Cases of Al at Kadaster

- 1. Fieldsketch vectorization
- 2. Detecting topographical elements
- 3. Disclosing cadastral information to the public



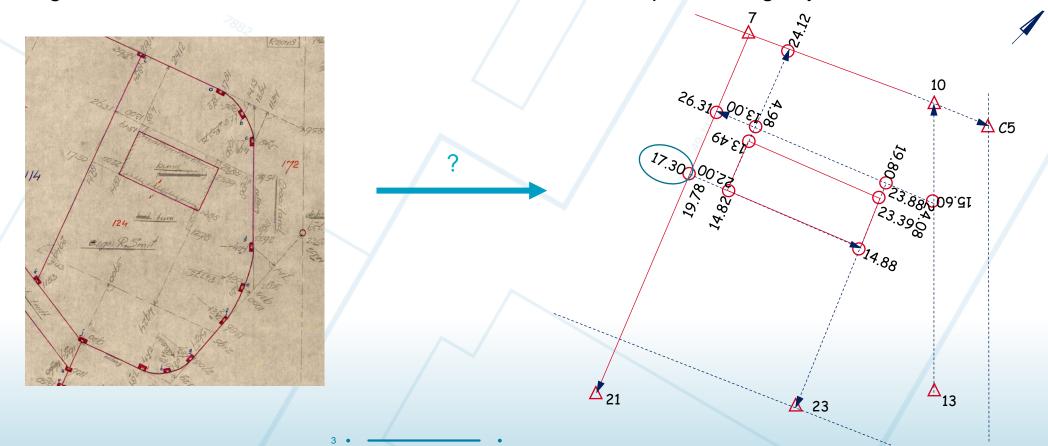




# 1. Fieldsketch Vectorization

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We are digitizing our archive of fieldsketches in order to achieve a more positive registry





# Algorithms for Vectorization: Lines

Detection of pixels belonging to one of the classes

- 1. Binary segmentation
- 2. U-NET based on Efficientnet-B5 architecture
- 3. From line pixels to vectorized lines

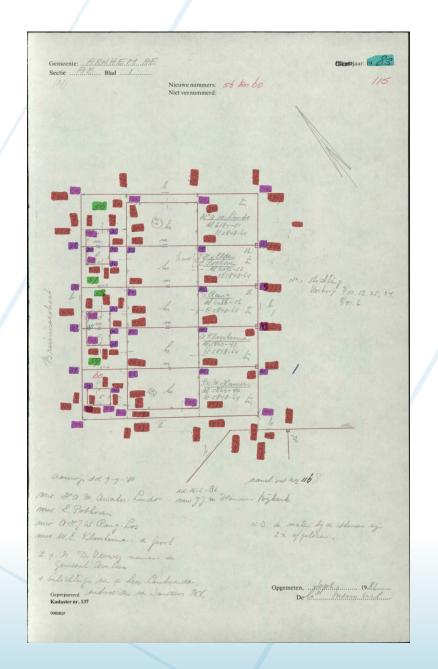
Buildings Lines Borders



# Algorithms for Vectorization: Objects

Detection of classes of objects on fieldsketches

- 1. Mask-RCNN using vision transformers backbone
- 2. Classes: measurements, parcel numbers, year etc
- 3. Linking of objects to lines

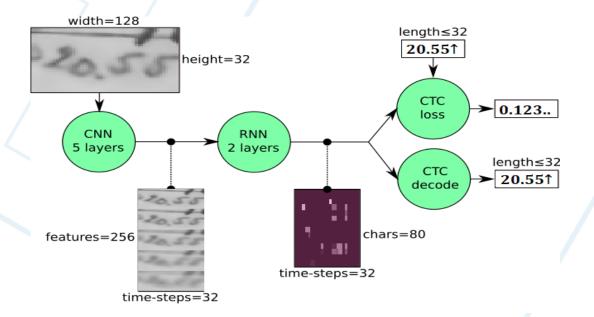




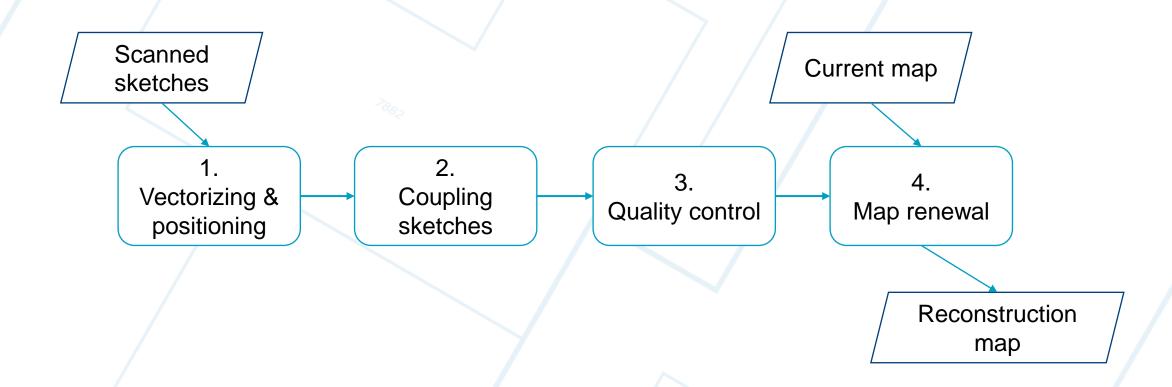
# Algorithms for Vectorization: Handwritten Tekst

#### Interpretation of objects

- 1. Reading handwritten tekst
- 2. Content of measurements, parcel numbers etc.
- 3. Neural Network with CNN and RNN layers
- 4. CTC loss merges output



# From Sketch to Map





# 2. Detecting Topographical Elements

We want to make object registration more efficient by using AI on image data

- 1. POCs on objects registred by the dutch cadastre
- 2. Detected objects are checked by humans





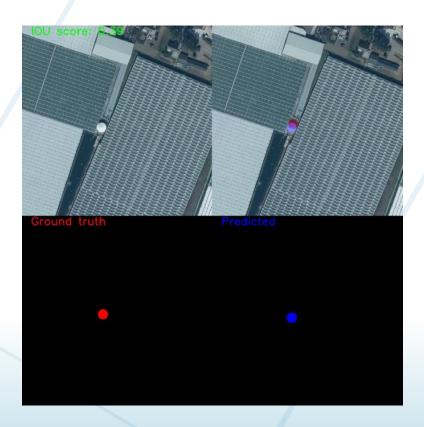
# **Ground Truth Data**

#### Challenges

- 1. Data availability
- 2. Data quality

#### **Solutions**

- 1. Manual labelling
- 2. Transfer learning
- 3. Data set augmentation





# Method

Multiple data sources (or a combination) may be used to detect objects on

- 1. Streetview (parking garages)
- 2. Oblique
- 3. Sattelite imagery (tanks)

Algorithm Used: U-Net







# Disclosing Land Registry Information to the Public

#### Goal:

Low-threshold access to Land Registry data based on natural language.

#### **Challenges:**

Easy access (multiple sources, complex data models, missing links)
Transparency and traceability (statistical AI methods, discrete explanation)

#### Techniques:

- 1. Knowledge Graph
- 2. Natural language Al





# **Question to Query Translation**

Any NLP task can be written as a text-to-text task.

#### Google translate task:

Converting a source language to a target language

#### Our task:

Converting a question to a SparQL or GraphQL query



# Method

Training a T5 language model and tokenizer to generate queries given input questions

#### **Examples:**

Q: How big is the garden of hofstraat 110 in Apeldoorn?

Q: Give me all buildings with construction year 2000 in Apeldoorn

External Q: Perform an inspection of hofstraat 110 in Apeldoorn in 10 minutes by bike

#### **Process:**

- 1. Convert query to tokens
- 2. Generate an output token sequence recursively based on the input sequence
- 3. Convert token sequence to query



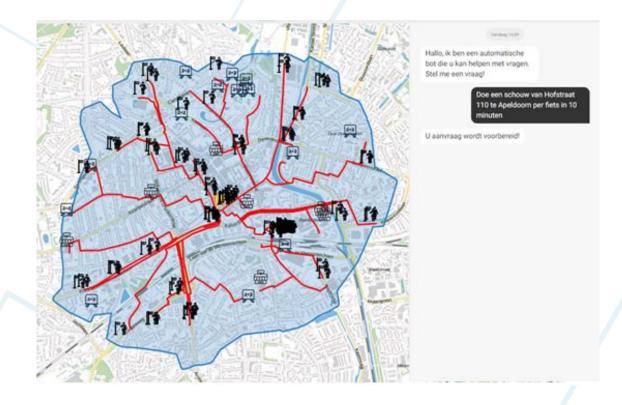
# **Current Status**

Integration with applications: API

- 1. Chatbot
- 2. Map viewers

#### Goal:

- 1. Geodataplein
- 2. Serve public cadastral information to citizens



# To Conclude

Al has a wide variety of applications within the Dutch cadastre

Human validation stays important

Al helps us to efficient retrieve, process and serve information

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