

Deutsches Forschungszentrum für Künstliche Intelligenz GmbH



What AI and Robotics could and couldn't offer for re-thinking agricultural systems

Joachim Hertzberg

Osnabrück University, Institute for Computer Science, and DFKI Labor Niedersachsen, Osnabrück, Plan-based robot control research group



2. ... and their potential for AgTech



2. ... and their potential for AgTech

Informatics methods and techniques for building artificial systems (agents) able to act in a goal-directed way in environments featuring incomplete control, independent dynamics, and/or imperfect knowledge.



Stuart Russell and Peter Norvig. Artificial Intelligence: A Modern Approach (4th Ed.). Pearson, 2020.

For example



Is # an x or an o? Decision based on incomplete information:

- model-based nominal (e.g., logics, rules, search)
- model-based statistical (e.g., Bayesian networks)
- data-based (e.g., machine learning methods)

For many good reasons, Machine Learning methods have had a great impact recently, but ...



Isn't the Pythagorean Theorem still math?







2. ... and their potential for AgTech

- machine learning, robotics, semantic technologies

Some projects in the background of this part

- Agro-Nordwest Experimentierfeld zur digitalen Transformation im landwirtschaftlichen Pflanzenbau (BMEL)
- Agri-Gaia Ein agrarwirtschaftliches KI-Ökosystem (BMWI)
- RESKIL Ressourceneffiziente KI für eingebettete Systeme in Landmaschinen (BMEL)
- CognitiveWeeding Selektives Unkraut- und Beikrautmanagement mit Hilfe Künstlicher Intelligenz (BMU, upcoming)





- Whenever you introduce new technology, consider re-thinking the old processes!
- You Ag people must say where to go, not we AI people!
- But do understand the potential of AI, to see what it could help in the re-thinking!

#1: machine learning/data analysis

8

(9)

 $\overline{7}$

(4)

e.g., weeding & volunteer potatoes, sugar beet classification, potato blight



(2)

Img: Kai Winkel (DFKI), ARGUS monitoring

OSNABRÜCI

Feature: how far will trained classifiers get?

- E.g., crop-weed-"other plant" distinction
- trained classifiers likely single chance of doing it
- challenges:
 - greatly different appearances (stadium, environment conditions, health state, ...)
 - get enough (classified) training images –
 synthetic images good enough for (pre-)training?
 - individual plants not segmented in situ
 - amalgamate appearance/data with expectation/knowledge about type and utility of non-crop plants



VS.





SWOT ML/data analysis in Ag

Strengths

 make use of rich data corpora available (machines/processes, geo-/satellite data, weather, ...) – no use running a farm without

Opportunities

- connect data across the complete agri-food value chain in digitalization (⇒ GAIA-X)
- create data-based business models in Ag

Weaknesses

- data quality often insufficient for training (lack of "clean", un-biased, classified data)
- selection of training data crucial for validity of training results
- using ML/DA technology currently impossible for most persons in Ag
- lack of integration of ML technology and knowledge-based methods

Threats

- "expectation failure" for data analysis methods/technology in Ag (as elsewhere)
- lack of acceptance for data transparency ("ownership", privacy issues, fear of being controlled, ...)

#2: "agricultural robotics"

Springer Handbook of Robotics (2016₂) Ch. 56: Robotics in Agriculture and Forestry (M. Bergerman, J. Billingsley, J. Reid, E. van Henten)

There is no such thing as an agricultural robot. There is robotics technology in agriculture.



Joachim Hertzberg/ 2021-06-28

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Feature: long-term robot autonomy ...



... requires context-sensitive action, requires a hybrid working area representation, including a semantic layer for closed-loop planning, monitoring, and user communication!



from sensor data & other available data

Kisliuk, B. & al., GIL-2021

SWOT robotics technologies in Ag



Strengths

- wide variety of SW and HW components integrated in closed-loop, with much full-/semi-automation potential
- robots often scalable in size and "team" size
- first exemplars/functions in the market

Weaknesses

- technology as a whole subject to research
- autonomous manipulation capabilities often lacking (hardware components brittle/clumsy)
- replacing current processes/components 1:1 by robotic technology makes little sense
 ⇒ re-thinking processes needed

Opportunities

 huge potential for economically and ecologically useful machines/processes in the context of transformation of Ag and digitalization of the agri-food chain

Threats

- lack of co-evolution of machinery and processes leads to lack of acceptability of technology
- lack of acceptance of technology among stakeholders or society
- lack of suitable legal & financial regulations for long-term autonomous physical machines

#3: semantic technologies in Ag

- UNIVERSITÄT OSNABRÜCK
- even if data can be analyzed and classified, semantic description layer needed for communicating results and for reasoning (machines, humans, nodes in value chains, ...)
- much-used AI approach: ontologies

 (e.g., in Semantic Web: OWL web ontology language)
- many approaches in Ag since a long time (AGROVOC/FAO, iGreen ontology/KTBL+DFKI a.m.o.)
- crucial description layer for system explainability, verifyability, certification

Feature: semantic process monitoring



Challenge: based on real telemetry+sensor data, monitor co-operative process (maize harvest) live(!) in terms of an agricultural ontology, including qualitative spatial & temporal relations



?sfh rdf:type agrico:Harvester ?sfh semap:hasObjectModel ?sfh_obj ?sfh_obj semap:hasLeftOfProjection2D ?sfh_proj_l2D ?tv rdf:type agrico:TransportVehicle ?tv semap:hasObjectModel ?tv_obj ?tv semap:hasConvexHull2D ?tv_abstr_ch2D ?tv_abstr_ch2D semap:isln2D ?sfh_proj_l2D >>

?tv agrico:positionedForLoading ?sfh



Arrival

Ready For

Loading

Deeken, H. & al., J. Appl Intell 2019

Departure

SWOT semantic technologies in Ag

Strengths

- sem. tech. offer description layer for describing, analyzing, communicating, reasoning on a high layer of "meaning"
- rich knowledge, high TRL regarding the resp. formalisms, technology

Weaknesses

- no widely accepted and used ontology in Ag
- interpreting data in terms of semantic categories is among the chief non-understood problems in AI in general ("Moravec's paradox")

Opportunities

- widely accepted and shared semantic layer for Ag applications would boost manufacturerindependent information exchange (⇒ GAIA-X)
- sem. tech. would help make Ag applications more robust (specification, verification)

Threats

 lack of acceptance of sem. tech. approaches (e.g., due to Moravec's paradox)





2. ... and their potential for AgTech

To take home



- AI has a lot of technologies to offer old ones and new ones
- Ag is a perfectly fitting application area for AI (data-rich, knowledge-rich, uncertain, under no full control, yet structured and goal-directed)
- Al goes well together with digitalization
- some AI-based components may improve existing functions or processes (some do it already); others would make sense only in widely re-structured processes
- in re-thinking Ag processes, Ag, not AI, persons (scholars, practitioners) must lead
- understanding basic AI principles is required for Ag persons in developing the right expectations wrt. using AI in the re-thought Ag systems
- the re-thinking appears to go well together with the general agri-food chain transformation that is requested by so many

Thank you for your time! Questions?







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