



Joint FAO/IAEA Centre Nuclear Techniques in Food and Agriculture

Al4Atoms Technical Meeting: Summary of Food and Agriculture Working Group Sessions

Session 1: On-line, 27 October 2021 15:40 – 18:30 (Vienna, CET) Session 2: On-line, 28 October 2021 15:40 – 17:55 (Vienna, CET)

Technical Officers: Mr Simon KELLY and Mr Gerd DERCON

Working Group Session 1: Programme



Day	Time	Duration	27-Oct	
27-10-2021	15:40	00:05	Working Group Session #1	
	20110		Food and Agriculture - Introduction	
	15:45	00:25	Frank Albinet (independent statistics and Al consultant, France)	
			Optimizing the remediation of radioactive contamination in agriculture using Deep Learning	
	16.10	00:25	Kate Kemsley (Quadram Institute Bioscience, United kingdom)	
	10.10		AI applied to NMR spectral data processing for food authentication	
	16:35	00:25	Yamine Bouzembrak (Wageningen Food Safety Research, Netherlands)	
			Machine learning applied to prediction of Food Fraud events	
	17:00	00:10	Comfort break	
	17.10	00.25	Modou Mbaye (Institut Senegalais de Recherche Agricole, Senegal)	
	17:10	00:25	Deep Learning Approach for Calibrating Cosmic-Ray Neutron Sensors (CRNS) in Area-Wide Soil Moisture Monitoring	
	17:35	00:25	Anne Gobin (VITO/KULeuven, Belgium)	
			Remote sensing data for agricultural soil management using machine learning algorithms	
	40.00		Panel Discussion (including Hans Marvin from Plenary Session)	
	18:00	00:30	Panel discussion and round-up of potential cross-cutting applications of AI and machine learning in the Agri-food space	
			End	

State of the Art (1)



Some lessons learnt from the keynotes – Food and Agriculture

- Enhanced data availability allowed to implement AI approaches
- Variation in sampling, sample preparation and analysis is a bottle neck for linking data, but AI can assist here to align datasets.
- Improving prediction
 - Chemometrics versus AI in MIRS NMR and other analytical techniques
 - Calibration of instruments
 - Calibration transfer (still limited experience)
 - Fingerprinting combined with AI (including stable isotope signatures)

State of the Art (2)



Some lessons learnt from the keynotes – Food and Agriculture

- Data fusion from different sources data access is a challenge
 - AI applications allowing linking point, local and regional/global information
- Sharing knowledge instead of sharing data
- Online open-access data mining for enforcement
- System approach combining Al and expert knowledge
- From prediction to explainability
- Internet of Things combined with AI and decision-support is a fastgrowing domain

Working Group Session 2:

Focusing on next steps, IAEA's role and expected outcome



Finding synergies across Food & Agriculture disciplines

- Data collection and ownership
- Data sharing (and ethics?)
- Protocols and verifying models (uncertainty)

Using AI in enforcement and for informing policy

Education and training

- Dissemination and best practice
- Real world examples and success stories
- Sustainability of the working group

Next steps (1) Data for AI applications in Food & Agriculture

- Limited data availability as compared to other scientific disciplines
 - Expensive and labour intensive data collation and annotation, and analysis (specialised devices and expertise needed)
 - ✓ Agronomic experiments often take years, so data collection is slow
- FAIR principle (findability, accessibility, interoperability, reusability of digital assets, with minimal human intervention)
 - A set of principles to ensure that the data are shared in a way that enables and enhances reuse by human and machines
 - ✓ Metadata may link the datasets through AI
- Federated learning as basis for sharing knowledge instead of sharing and moving data (training the model from one database to another database)
- Al can assist in calibration transfer for analytical instruments, essential to develop large uniform datasets (e.g. spectral analysis)
- ✓ Computing power may no longer be a major limitation
- ✓ AI based digital twins allow to explore processes in virtual reality (based on gaming applications)
- Data fusion and integration are a challenge due to the wide range of data types (from farm to fork, from point to region)

Next steps (2)

Using AI in explaining processes, informing policy and enforcement

✓ Explainable AI

- ✓ Explainability should not be a limitation for deployment of AI techniques
- ✓ But limited explainability may be needed to make AI attractive to enforcement or policy makers and to make sure that the end-user understands it in their environment.

✓ Explainability of processes through AI is more for the domain experts, not so much for the end-user.

- ✓ However, main drivers for a certain result (classification) should be explainable for the end-user (so not just black box).
- ✓ Algorithms may be not inclusive enough, so that it may be a source for certain bias (how to overcome?).
 - ✓ This is linked to the ethical questions. And how can we use such models?
- Reactive AI modelling is embedded in the legal framework (required to be done), and proactive AI modelling is not embedded in the legal framework (if not happened yet, how to justify to take action to prevent it).
- Can AI approaches, such as data mining, also be a basis for keeping track of rapidly increasing amount of information (online/in papers)?
 - ✓ It is already being done, such as for crop yield data, prediction of zoonotic diseases, impact of climate change on diseases (shell fish), ...

Next steps (3) Education and training

✓ AI started to be included in curricula of universities and faculties for agricultural sciences, mainly at PhD, but also at MSc level.

- ✓ In Wageningen included in MSc courses (different research programmes working together on AI, including teaching, 3 million Euro per year available for AI research, with four chairs).
- ✓ In KULeuven computer science and engineering driven, with some introduction in the field in bio-engineering (more at PhD level).
- ✓ Sufficient information available online for implementing AI (excellent courses available)
- ✓ But platform would be needed to guide students.
- ✓ Theoretical and practical AI are often separate fields.
 - ✓ For instance, MIRS AI applications are carried out by soil scientists, and not mathematicians.
 - ✓ More **interaction** is needed.
- Not just universities but also high schools (secondary schools) have a role to play in education in the field of AI (to ensure capacity and interest later onwards).

✓ Gaming can be a basis.

Accelerating Progress – IAEA's role

- ✓ Through Coordinated Research Projects, the Joint FAO/IAEA Centre develops new AI approaches.
- ✓ Through TC projects and IAEA capacity building programmes, know how on AI applications can be disseminated to Member States.
- ✓ Progress is stimulated by interdisciplinary exchange of expertise and experiences across scientific fields (across different departments of the IAEA)
 - Theory versus application
- The FAO/IAEA AI Working Group decided to meet once a year to exchange experience on approaches, data exchange, and project development.
 Let us know if you are interested in joining us.

Expected Outcome

- Al helps to fuse and integrate data and datasets
 - From local to global datasets
- AI innovates model development for enhanced decision support and enforcement in a scientific and ethical way (based on FAIR principles)
- Al becomes a mainstream tool for better use of nuclear and isotope data
- Al is **integrated in education programmes** at different levels

Al for Good

- Based on the Al4Atoms TM Food and Agriculture Working Group outcomes, one of our speakers Franck Albinet, will have the opportunity to speak at the "Al for Good" <u>https://aiforgood.itu.int</u> on 18-11-2021; a major initiative organized by the ITU with 38 UN sister agencies
- Welcome to this event! And you can register already.

Thank you for your participation in the Food and Agriculture Working Group Sessions



Joint FAO/IAEA Centre Nuclear Techniques in Food and Agriculture