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LIHLITH: Improving Communication Skills of Robots Through Lifelong Learning

by Eneko Agirre, University of the Basque Country (UPV/EHU), Sarah Marchand, Synapse Développement, France, Sophie Rosset, Computer Science Laboratory for Mechanics and Engineering Sciences (LIMSI), Anselmo Peñas, Université Nationale de l’Éducation à Distance (UNED) and Mark Cieliebak, Zurich University of Applied Sciences (ZHAW)

Dialogue systems are a crucial component when robots have to interact with humans in natural language. In order to improve these interactions over time, the system needs to be able to learn from its experience, its mistakes and the user's feedback. This process – fittingly called lifelong learning – is the focus of LIHLITH, an EU project funded by CHIST-ERA.

Artificial Intelligence is a field that is progressing rapidly in many areas, including dialogues with machines and robots. Examples include speaking to a gadget to request simple tasks like turning on the radio or asking for the weather, but also more complex settings where the machine calls a restaurant to make a reservation [1], or where a robot assists customers in a shop. LIHLITH [2] is a project focusing on human-machine dialogues. It aims to improve the self-learning capabilities of an artificial intelligence. More specifically, LIHLITH will devise dialogue systems which learn to improve themselves based on their interactions with humans.

LIHLITH (“Learning to Interact with Humans by Lifelong Interaction with Humans”) is a three-year high risk / high impact project funded by CHIST-ERA [3] that started in January 2018. Participating partners are researchers from University of the Basque Country (UPV/EHU), Computer Science Laboratory for Mechanics and Engineering Sciences (LIMSI), Université Nationale de l’Éducation à Distance in Spain (UNED), Zurich University of Applied Sciences (ZHAW), and Synapse Développement in France.

Current industrial chatbots are based on rules which need to be hand-crafted carefully for each domain of application. Alternatively, systems based on machine learning use manually annotated data from the domain to train the dialogue system. In both cases, producing rules or training data for each dialogue domain is very time consuming, and limits the quality and widespread adoption of chatbots. In addition, companies need to monitor the performance of the dialogue system after being deployed, and re-engineer it to respond to user needs. Throughout the project, LIHLITH will explore the paradigm of life-long learning in human-machine dialogue systems with the aim of improving their quality, lowering the cost of maintenance, and reducing efforts for deployment in new domains.

Main goal: continuous improvement of dialogue systems
The main goal of life-long learning systems [4] is to continue to learn after being deployed. In
the case of LIHLITH, the dialogue system will be developed as usual, but it will include
machinery to continue to improve its capabilities based on its interaction with users. The key
idea is that dialogues will be designed to get feedback from users, while the system will be
designed to learn from this continuous feedback. This will allow the system to keep
improving during its lifetime, quickly adapting to domain shifts that occur after deployment.

LIHLITH will focus on goal-driven question-answering dialogues, where the user has an
information need and the system will try to satisfy this need as it chats with the user. The
project has been structured in three research areas: lifelong learning for dialogue; lifelong
learning for knowledge induction and question answering; and evaluation of dialogue
improvement. All modules will be designed to learn from available feedback using deep
learning techniques.

Figure 1 shows, in white boxes, a standard dialogue system. These systems use natural
language understanding to process the user input, dialogue management to access the
domain knowledge and decide the next utterance, and natural language generation to
produce the next utterance of the system.

The lifelong learning module, in blue, is the key innovation of LIHLITH, and will improve all
modules as the system interacts with users, updating the domain knowledge accordingly.
The project will explore autonomous reconfiguration of dialogue strategies and proactive
capabilities to ask the user for new knowledge and for performance feedback. This will be
triggered, for instance, when the past reactions have been rejected, when the user
interaction is too ambiguous, when the possible answers are too numerous, or if they have
too similar confidence scores.

To carry out this research, LIHLITH combines machine learning, knowledge representation
and linguistic expertise. The project will build on recent advances in a number of research
disciplines, including natural language processing, deep learning, knowledge induction,
reinforcement learning and dialogue evaluation, to explore their applicability to lifelong
learning.

Links:

References

Please contact:
Eneko Agirre
University of the Basque Country (UPV/EHU)
e.agirre@ehu.eus
Figure 1: Schema of a standard dialogue system in white boxes. The innovative lifelong learning module is able to improve all modules (in blue) based on past interactions and the interaction with the current user, updating the domain knowledge accordingly.