



**Programme: Intelligent Systems** 

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Who am I

## Who am I



## **Background**

- Software Engineering BSc
- Artificial Intelligence MSc

#### Where to find me

- Resumé
- Google Scholar

## Currently

- International project LIHLITH-KIQA (PI: Anselmo Peñas).
   Question Answering that evolve over time.
- First year PhD in Lifelong Learning Question Answering.

# Introduction

## **Introduction: Lifelong Learning**



## **Lifelong Learning**

Framework to study how to enable systems to evolve over time, integrating new knowledge with previous one and adapting to new tasks.

Can be applied to multiple learning paradigms and techniques:

- Active Learning
- Reinforcement Learning
- Transfer Learning
- Knowledge acquisition

# **Introduction: Question Answering**



## **Question Answering**

Systems that automatically answer questions posed by humans in a natural language

#### **Focus**

- Based on Knowledge Graphs
- Based on other resources: Set of documents, search engines

## **Applications**

 Question Answering Systems are everywhere, from personal assistants to chatbots and IT ticket management.

## **Thesis Frame**



## Most QA systems:

- Work under a closed world assumption, ranking plausible answers
- Learn only during training, becoming stale over time
- Fail to sustain performance in production environments

We focus on Lifelong Learning Question Answering systems, that:

- Evolve over time
- Integrate new knowledge with previous one
- Adapt to new tasks

# Research Plan

## Research Plan: Objectives



## **Objectives**

- Develop QA systems sustainable in production environments
- Develop QA systems capable of detecting knowledge gaps
- Develop strategies to fill these gaps
- Enable QA systems to evolve over time
- Enable QA systems to adapt to new tasks.

# Research Plan: Open Questions



#### **Research Questions**

- How to assess QA systems effectiveness
- How to adapt QA systems to new tasks
- How to detect unanswerable questions given the available knowledge
- How to incorporate new knowledge with previous one
- How machines and humans performance differ



## Goal: Study the state-of-the-art

Related to Knowledge Graphs:
 Guillermo Echegoyen, Álvaro Rodrigo, Anselmo Peñas (2019).

 Benchmarking Entity Linking for Question Answering over Knowledge Graphs. Sociedad Española de Procesamiento de Lenguaje Natural (PLN). Volume 63, pages 121-128. ISSN: 19897553, 11355948. DOI: 10.26342/2019-63-13. [2]

#### **Outcomes**

- 6 Entity Linking evaluation datasets.
- EL has a higher impact over QA systems than usually though.



## Goal: Study the state-of-the-art

Related to NLP and Dialogue Systems:
 Jan Deriu, Alvaro Rodrigo, Arantxa Otegi, Guillermo Echegoyen,
 Sophie Rosset, Eneko Agirre, Mark Cieliebak (2020). Survey on evaluation methods for dialogue systems. Artificial Intelligence Review. DOI: 10.1007/s10462-020-09866-x. [1]

#### **Outcomes**

- Trend towards end-to-end, more cryptic systems, based on large amounts of data.
- Trend towards less human involvement regarding evaluation.



## Goal: Study the state-of-the-art

Related to Lifelong Learning (WIP):
 Guillermo Echegoyen, Álvaro Rodrigo, Anselmo Peñas (2019).
 Study of a Lifelong Learning Scenario for Question Answering

#### **Outcomes**

- Transfer learning between different extractive QA datasets works.
- Catastrophic forgetting is a big concern.



## Goal: Study the state-of-the-art

Position papers (shared task proposal):
 Anselmo Peñas, Mathilde Veron, Camille Pradel, Arantxa Otegi,
 Guillermo Echegoyen and Alvaro Rodrigo (2019). Continuous
 Learning for Question Answering. Dialogue Systems and
 Lifelong Learning. Increasing Naturalness and Flexibility in Spoken
 Dialogue Interaction: 10th International Workshop on Spoken
 Dialogue Systems, Lecture Notes in Electrical Engineering, Springer.
 ISSN: 1876-1100 [4]

#### **Outcomes**

 Task: Determine mappings from utterances to a KG and means to enrich the KG.



## Goal: Study the state-of-the-art

Position papers (LL Evaluation collection):
 Mathilde Veron, Anselmo Peñas, Guillermo Echegoyen, Somnath
 Banerjee, Sahar Ghannay, and Sophie Rosset (2020). A Cooking

 Knowledge Graph and Benchmark for Question Answering

 Evaluation in Lifelong Learning scenarios. Natural Language in
 the Database and Information Systems (NLDB), Lecture Notes in
 Computer Science, Springer 2020. Volume 12089, pags. 94-101.
 ISSN: 0302-9743 [5]

## Outputs

- Cooking Domain Knowledge graph
- Cooking Domain Question Answering dataset



## Goal: Study the state-of-the-art

Lifelong Learning and transfer learning:
 Guillermo Echegoyen, Álvaro Rodrigo, Anselmo Peñas (en prensa).
 Cross-lingual Training for Multiple-Choice Question
 Answering. Sociedad Española de Procesamiento de Lenguaje
 Natural (PLN). Accepted in March 2020, Volume 64, September.
 ISSN: 1989-7553. [3]

## Outputs

- Exams dataset: Difficulty affects human and machines in a similar way.
- QA systems can be transferred to different tasks with acceptable performance, even in different languages.



# Assessing a lack of knowledge is key

- Propose method to address unanswerable Questions
- Propose method to incorporate new knowledge
- Craft evaluation collections for the task

# Transfer Learning between languages & tasks

- Cross-lingual models
- Multiple Reading Comprehension datasets



# QA System capable of:

- Learning new tasks without forgetting previous ones
- Assessing what it doesn't know and trigger learning procedures
- Integrating new knowledge with previous one
- Adapting to new tasks

# **Conclusions**

## **First Year Conclusions**



# Experimentation is hard

- Money for value
- Interpretability is not always clear
- Catastrophic forgetting is hard to overcome
- We are far from real world QA systems
- Ideal laboratory settings are far from real world use cases

## **First Year Conclusions**



# Trend towards Deep Learning

- Large code base
- Programming GPUs/TPUs
- Experiments are costly
- Cloud computing
- Models are not interpretable
- Research grows exponentially

# Thank you! Questions?

# Guillermo Echegoyen

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#### References i



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Survey on evaluation methods for dialogue systems.

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Procesamiento del lenguaje natural, (63):121–128, 2019.





G. Echegoyen, Á. Rodrigo Yuste, and A. Peñas Padilla.

Cross-lingual Training for Multiple-Choice Question Answering.

2020.



A. Peñas, M. Veron, C. Pradel, A. Otegi, G. Echegoyen, and A. Rodrigo.

Continuous learning for question answering.

In Tenth International Workshop on Spoken Dialogue Systems Technology (IWSDS), 2019.

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M. Veron, A. Peñas, G. Echegoyen, S. Banerjee, S. Ghannay, and S. Rosset.

A Cooking Knowledge Graph and Benchmark for Question Answering Evaluation in Lifelong Learning Scenarios.

In E. Métais, F. Meziane, H. Horacek, and P. Cimiano, editors, Natural Language Processing and Information Systems - 25th International Conference on Applications of Natural Language to Information Systems, {NLDB} 2020, Saarbrücken, Germany, June 24-26, 2020, Proceedings, volume 12089 of Lecture Notes in Computer Science, pages 94–101. Springer, 2020.