NLP-chatbot R&D at CDS. VO standards, Al impact, etc.

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Purpose

- Feedback and status of a long term R&D action at CDS
- A study around Natural Language Processing applied to the astronomical data services querying
- It was presented step by step in various IVOA groups
- Taking now into account the Al vawe...

□ Why NLP?

• Since Web advent we are used to deal with webpages containing forms (parameter fields, checkboxes, etc.)



- Nowadays we use the voice (gesture should join for multimodal requests) to interact with several devices with voice answer or a displaying
- An interesting field of investigation with both technical and human concerns

Webpages / Forms pro cons

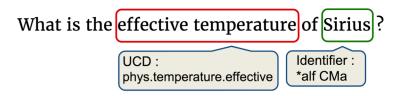
- Forms are not (yet) old fashion and have still advantages
 - Easy and quick for experimented users
 - easy to translate to database queries
 - Result rendering "easy" to design
- On the negative side :
 - not intuitive for new or occasional users
 - Thus we use often only a (small) part of the functionalities
- Understanding / Knowledge / Learning efforts on the user side

NLP pro cons

- A friendly (Inclusive if disabilities) way of interaction
 - More digestible than forms
 - Just a prompt or a voice recognition
 - Result displaying can be similar than for forms or a speech synthesis
- On the negative side :
 - possible frustration for expert users if the result is too imprecise
 - too long chat if the question was vague or ambiguous
- User request Understanding and translation to query the services is on the service side

Pragmatic and best effort approach

- Learning about NLP (basis, tools, examples, ...)
- Starting by building a set of typical queries
- We had already a lot of helpful data and tools
 - authors in Simbad, VizieR
 - missions and wavelengths in VizieR
 - DJIN to recognize identifiers in a text
 - UCDs
 - a name resolver



And a useful friend : the Virtual Observatory

- We used VO standards like TAP/ADQL (DAL WG), UCDs (Semantics WG), ..., and tools based on ML (and now AI), a link with KDD IG.
- The VO was a master piece which helped us to translate NL into queries to our own services. By the way all services implementing VO standards could be included.

```
What is the effective temperature of Sirius ?

List the QSOs at Z > 6. Simbad, TAP query SELECT "VI/137/gum_mw".Teff VizieR, TAI FROM "VI/137/gum_mw"
```

```
SELECT main_id, oid, rvz_redshift
FROM basic
WHERE otype = -14680064 AND rvz_redshift > 6;
```

```
SELECT "VI/137/gum_mw".Teff VizieR, TAP Query FROM "VI/137/gum_mw" WHERE 1 = CONTAINS(POINT('ICRS', "VI/137/gum_mw"."RAJ2000", "VI/137gum_mw"."DEJ2000"), CIRCLE('ICRS', 101.287155333, 16.716115861, 20/3600.));
```

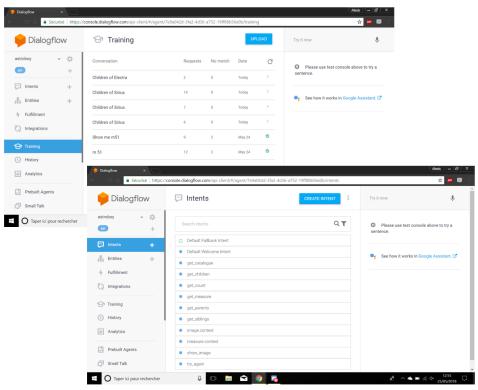
On the way to a chatbot

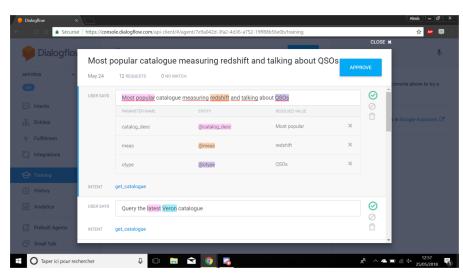
- A first work with the Stanford NLP library was far from a Chatbot as it must be able to chat with the user otherwise it is just like a command line
- Google Dialogflow brought us this chat capability
 - Based on Machine Learning
 - Training, tuning, improving features



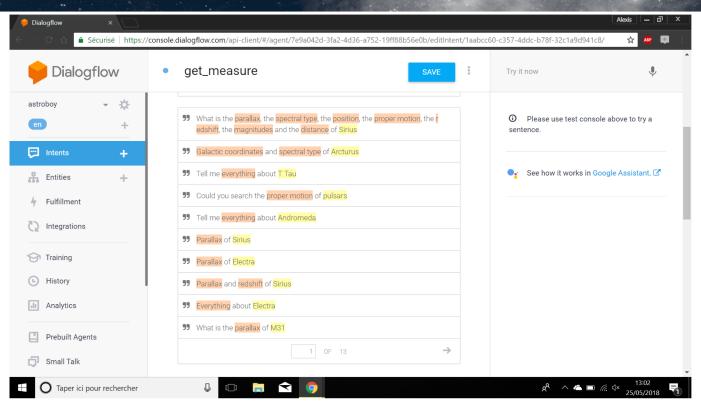
Query understanding was then delegated to Dialogflow

Switching from coding to training to understand user intents





Examples of sentences linked to get_measure



Remark: External tools

- With Dialogflow we delegated a larger part of the tool than with the first approach but not the link to the services.
- The link to the services and de facto to the data is the key part and our development was always designed to switch to other NLP tools.
- We switched to an Open Source framework, Rasa



How does Rasa work ?

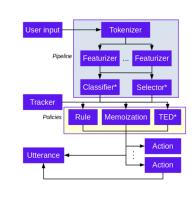
... Rasa prediction model training is based on Tensorflow

Compared to Dialogflow the configuration is not really user-

friendly

Items of the architecture have to be written

For example a CDSTokeniser



language: "en'

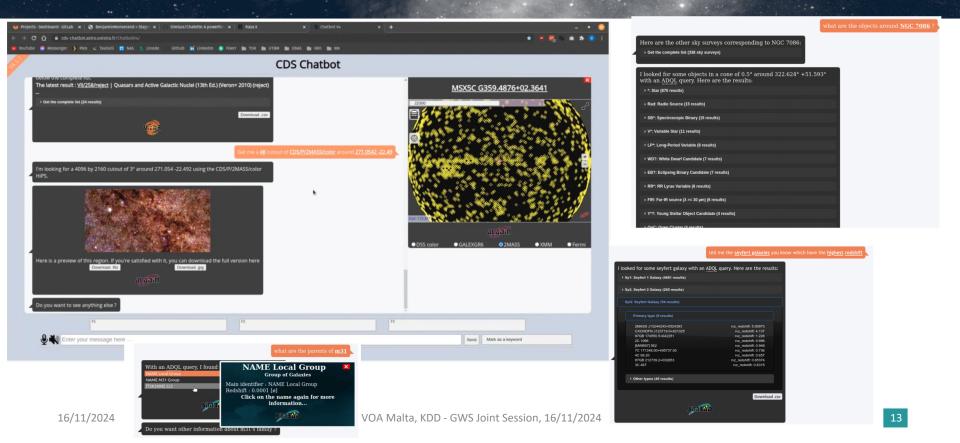
pipeline:

- name: RegexFeaturizer
- name: LexicalSyntacticFeaturizer
- analyzer: "word"
- analyzer: "char_wb" min_ngram: 1
- max_ngram: 4
- name: DIETClassifier epochs: 100
- name: ResponseSelector epochs: 100

policies:

- name: MemoizationPolicy max history: 5
- name: **TEDPolicy** max_history: 5
- epochs: 200 - name: RulePolicy

Current version based on Rasa framework



The Day after

- Work with Rasa was continued (cleaning, refactoring, etc.) but with the advent of chatGPT we decided to make a break and to spend time on the study of its impact, a help for the existing chatbot? or a way to reshape it completely?
- We had a look at several AI tools and we decided to test with the OpenAI API which cost model (credits to buy and not a monthly subscription) is possible for us

First Experiments with OpenAl

- OpenAI has a functionnality called assistant which allows the definition of assistants inside an application, able to answser a request using models, tools and data specified by the developper
- A first prototype was defined to learn Simbad table structure and ADQL to become able to propose or verify TAP queries to Simbad.

Examples

User

give me the TAP request of the position of Vega

TAP Query

To obtain the TAP (Table Access Protocol) request for the position of Vega, I will use the "basic" table, which contains general data about astronomical objects, including their coordinates (right ascension and declination) [7*source]. The TAP query will involve selecting the right ascension (RA) and declination (Dec) for the object identified as Vega.

Let me construct the TAP query for you.

TAP Query

The TAP query to retrieve the position of Vega from the Simbad database would be:

```
1 SELECT ra, dec
2 FROM basic
3 WHERE main_id = 'Vega'
```

This query selects the right ascension (ra) and declination (dec) from the "basic" table, where the main i (main_id) is equal to 'Vega'.

User

this not work because vega is not the main id. You must join with the indentifiers table

TAP Query

You are absolutely right. To obtain the position of Vega, I will need to join the "basic" table with the "ident" table using the object internal identifier (oidref).

The TAP query to retrieve the position of Vega from the Simbad database with the necessary join would be:

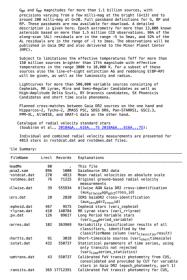
```
1 SELECT b.ra, b.dec
2 FROM basic AS b
3 JOIN ident AS i ON b.oid = i.oidref
4 WHERE i.id = 'Vega'
```

This query selects the right ascension (ra) and declination (dec) from the "basic" table and joins it with the "ident" table on the object internal identifier, then filters the result to only include the entry with the identifier 'Vega' in the "ident" table.

Ongoing experiments

 We have a mine of information not directly used in the queries to the services like the VizieR Readme files





LLMs are efficient for textual content The aim is to train with this additionnal data to detect the astronomical object names in the Readme files

Files of valid, invalid names to train, etc.

remark

- Testing / working with AI frameworks is often nearer to cooking than to programming
- best recipe to find
- Trying with different models
- Training with a good set of data (also a set with errors)
- Improving with other sources or various data formats for a better understanding by the models
- Etc.

Conclusion

- We are testing and improving step by step to interact with data services differently, keeping in mind the gap between a prototype and a production service
- .. And adapting us to new tools/frameworks/paradigms; so as usual in IT

Providing stable and sustainable new tools/services benefiting from AI is an interesting challenge

Questions?