NLP course 2021

Homework 3

WSD of Word-in-Context Data

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WSD of Word-in-Context Datasets



WiC is the task of addressing the disambiguation of polysemous words, without relying on a fixed inventory of word senses.



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- Use the mouse to click on the button
- The cat eats the mouse



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- Use the mouse to click on the button
- The cat eats the **mouse**

In the above examples, the word **mouse** has different meanings, because it is used in different contexts.



WiC is the task of addressing the disambiguation of polysemous words, without relying on a fixed inventory of word senses.

- The cat eats the mouse
- the **mouse** escaped from the predator

In the above examples, the word **mouse** has different meanings, because it is used in different contexts.



WiC: context sentences, target words

The mouse escaped from the predator



WiC: context sentences, target words

The mouse escaped from the predator

context sentence



WiC: context sentences, target words

The mouse escaped from the predator context sentence



The homework

WSD of Word-in-Context Data





The goal(s)

The task is formulated as a standard **Word Sense Disambiguation task**:

- Given each context sentence, determine the meaning of the each of the two occurrences of the target word.
- Evaluate the system in the Word-in-Context task too



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The cat eats the mouse _____ model ____



The goal

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- Evaluate the system in the Word-in-Context task too

Click the right button of your mouse ____ model ____



WSD: evaluation

The performance of your WSD system is usually measured in terms of F1 score

GOLD	True	False	True	True	True	False
PRED	True	True	True	True	False	False

F1 = 2PR / P+R = Accuracy if all answers are given



WiC: evaluation

The performance of a WiC system is usually measured in terms of accuracy.

GOLD	True	False	True	True	True	False
PRED	True	True	True	True	False	False

Accuracy = (CORRECTLY classified sentence pairs) / (TOTAL pairs) = 4/6 ~= 0.66



Evaluation Output

- For each sentence pair, we expect you to output the synset identifier for each sense following the input format
- You will get the WiC prediction "for free" as a result of WSD (i.e. a True/False if senses are the same or different for the two occurrences of the target word)



Practical tips

• Use NLTK WordNet API to obtain the synset information from the synset identifier, as well as the other synsets for that lemma:

```
Obtain the sense using the synset key:
    sense = wordnet.lemma_from_key("filthy%5:00:dirty:01")

Obtain the synset:
    synset = sense.synset()

Obtain the lemmas of a given synset:
    all_senses = wordnet.lemmas(lemma, pos)
    all_senses = wordnet.lemmas(sense.name(), sense.synset().pos())
```

Submission

What you will receive & How to submit



Data for WSD

- 1) http://lcl.uniroma1.it/wsdeval contains training data (SemCor) and benchmarks (SemEval, Senseval, etc.)
- 2) https://github.com/SapienzaNLP/ewiser/tree/master/res/corpora/training/origrecontains additional training data annotated with:
 - a) examples
 - b) glosses



What you will receive: data format for SemCor/Senseval data you will use for training

```
<sentence id="d0000001.s001">
      <wf lemma="it" pos="PRON">it</wf>
      <wf lemma="be" pos="VERB">was</wf>
      <wf lemma="full" pos="ADJ">full</wf>
      <wf lemma="of" pos="ADP">of</wf>
      <wf lemma="racket" pos="NOUN">rackets</wf>
     <wf lemma="," pos=".">,</wf>
      <wf lemma="ball" pos="NOUN">balls</wf>
      <wf lemma="and" pos="CONJ">and</wf>
      <wf lemma="other" pos="ADJ">other</wf>
      <instance id="d0000001.s001.h001" lemma="object" pos="NOUN">objects</instance>
     <wf lemma="." pos="PUNCT">.</wf>
 </sentence>
```



```
"id": "dev.24",
"lemma": "filthy",
"pos": "ADJ",
"sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements.",
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
"start1": "26",
"end1": "32",
"start2": "21",
"end2": "27",
"label": "True",
"sent1_sense_id": "wic.dev.s24.w01",
"sent2_sense_id": "wic.dev.s24.w02"
```

```
"id": "dev.24".
"lemma": "filthy",
                                   Input context sentences
"pos": "ADJ",
sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements."
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
start1": "26",
"end1": "32".
"start2": "21".
"end2": "27",
"label": "True",
"sent1_sense_id": "wic.dev.s24.w01",
"sent2 sense id": "wic.dev.s24.w02"
```

```
"id": "dev.24",
"lemma": "filthy",
"pos": "ADJ",
"sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements.",
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
                 First char (included) and last char (excluded)
                 of target word in sentence 1
"start2": "21",
"end2": "27",
"label": "True",
"sent1_sense_id": "wic.dev.s24.w01",
"sent2_sense_id": "wic.dev.s24.w02"
```

```
"id": "dev.24",
"lemma": "filthy",
"pos": "ADJ",
"sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements.",
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
"start1": "26",
"end1": "32",
                 First char (included) and last char (excluded)
                 of target word in sentence 2
"sent1_sense_id": "wic.dev.s24.w01",
"sent2_sense_id": "wic.dev.s24.w02"
```

```
"id": "dev.24",
"lemma": "filthy",
"pos": "ADJ",
"sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements.",
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
"start1": "26",
"end1": "32",
"start2": "21",
"end2": "27",
"label": "True", Gold label of this sample
"sent1_sense_id": "wic.dev.s24.w01",
"sent2_sense_id": "wic.dev.s24.w02"
```

```
"id": "dev.24",
"lemma": "filthy",
"pos": "ADJ",
"sentence1": "They were given food in a filthy toilet without even minimal sanitary arrangements.",
"sentence2": "He spent a week in a filthy cell together with seven other detainees.",
"start1": "26",
"end1": "32",
"start2": "21",
"end2": "27",
                                         This is the instance Id. They are used to
"label": "True",
                                         indicate the prediction in the output file.
sent1_sense_id": "wic.dev.s24.w01",
"sent2_sense_id": "wic.dev.s24.w02"
```

Output format

TSV file with:

```
wic.dev.s24.w01 filthy%5:00:dirty:01
wic.dev.s24.w02 filthy%5:00:dirty:01
wic.train.s02.w01 ...
```

Predictions for dev set sample 24. If the two senses are different, it will show a different synset, i.e.:

filthy%5:00:dirty:02



We will provide you a folder with the following structure:

```
nlp2021-hw3/
     data/
     hw3/
          model.py
      stud/
     model/
     requirements.txt
     test.sh
```

You are allowed to edit **only** the items in bold.

We will use Docker for evaluation. As far as you **do not change** any file but those we marked in bold, **if test.sh runs** on your side, it will run on ours as well.

What we expect from you

- The zip folder we gave you (but populated :))
- Put your training code (if you used Colab, download the notebook .ipynb and place it) in hw3/stud/
- If you use any additional library, modify the requirements.txt file as needed (click <u>here</u> for info)
- Use the data (train, dev and test) in the data folder; use each file as defined in the standard ML conventions (train for training, dev for model selection, ...)
- Put everything your model needs (vocabulary, weights, ...) inside the model/ folder, and be sure to properly load them in your model

What we expect from you

- 1. In hw3/stud/implementation.py implement the StudentModel class
 - Load your model and use it in the predict method
 - You must respect the signature of the predict method
 - You can add other methods (i.e. the constructor)
- 2. In hw1/stud/implementation.pyimplement the build_model function, initializing your StudentModel class.
- 3. Use test.sh to check that everything works
- 4. Add your report.pdf to the folder (yes, export it in pdf even if you are using Word!)
- 5. Name the zip folder lastname_studentid_hw1.zip
 - Ex: Luigi D'Andrea will submit a file named dandrea_1234567_hw1.zip

Submission instructions



- Upload the zip on your **institutional** Drive and make it **link-shareable** and **public to anyone** (an automatic script will download it).
- Make sure it is accessible via an incognito page of your browser!
- Do NOT modify the folder structure
- You have to submit the homework through the submission form on Google Classroom. You will be asked to fill a form with the requested information and the link to the zip you uploaded on Drive.

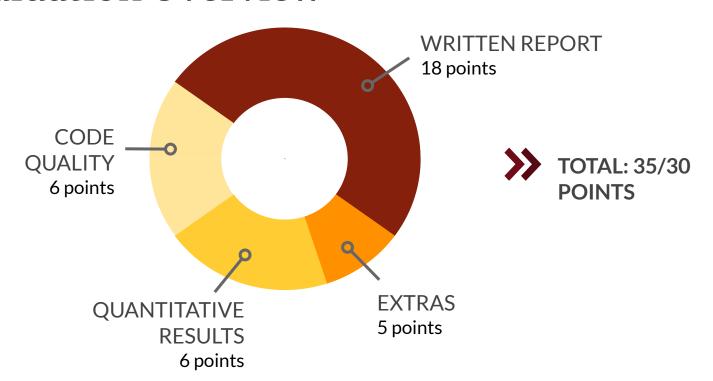
Evaluation

How your work will be evaluated





Evaluation Overview



Report: dos and don'ts

- ACL 2021 paper template
 - Available <u>here</u> (Word and LaTeX direct download) or <u>here</u> (Overleaf LaTeX template)
 - You can use either the LaTeX or the Word template, your choice
 - DO NOT MODIFY the template (margins, spacing, font size)
 - Use the non-anonymous flag, so you can enter your name
- Max 2 pages
 - For the report, including title, subtitles, etc.
 - This is a STRICT RULE!
- Unlimited extra pages for images, tables and references
 - Every image and table must have a caption (don't abuse them please :-))
 - Tables and images must be referenced in the report

Report: what you are expected to do

We expect a good report to be:

Readable and understandable

 We will not give penalties for English errors, but we expect the report to follow a clear flow. We don't want to read just a sequence of statements on what you did without showing the reasoning behind your choices

Well-structured and organized

 Take inspiration from the many papers available online and organize your report in well-defined sections (e.g. method, setup, experiments, results...)

Report: what you are not expected to do

We expect a good report **NOT** to include:

- Unnecessary task or dataset descriptions
 - just focus on your solution to the problem.
- Code copy-paste
 - Your code should be self-explanatory, so no need to show it in the report. You
 can add pseudo-code to show some particular algorithm, but no code or
 screenshots please!



Report: what you are not expected to do

We expect a good report **NOT** to include:

- Unnecessary low-level implementation details
 - Avoid any low-level implementation/technical details like "I used a dictionary to store these values", "I had to use configuration X to solve this exception", "I could not use Y because there was a dependency issue with Z", etc.
 - o Instead, we are interested in high-level abstractions/strategies you decide to use to tackle the homework, as well as the intuitions behind your choices.
 - E.g. use and description of a particular model, explanation of how and why an architecture works, etc.

Code and code Quality

Your project should conform to the following rules:

- You must MUST use PyTorch Lightning
 - TensorFlow and other deep learning frameworks are NOT allowed.
- Any additional framework is allowed (but needs to be documented in the report)
- Use of sense embeddings is allowed
- For any doubt, please ask the TAs on Google Classroom.
- Comment your code, please!



Quantitative Results

We will evaluate the **performance of your model** on a SECRET test set.

You can get from 0 to 6 points according to the following thresholds:



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You can get from 0 to 6 points according to the following thresholds:

Thresholds will be defined based on an internal reference model and the normalized distribution of YOUR scores!



Extras

You can achieve up to 5 points with some extras!

An "extra" is whatever you decide to add to your model to make it better. For instance:

- use of pre-trained embeddings or contextualized embeddings,
- use of NLP best practices,
- comparative analysis of results in your report,
- informative plots in your report,
- new ideas

and more, according to internal baselines. Don't forget to **explain your choices** in the report! Extras that are not explained in the report will not be considered for evaluation.

Evaluation

- test.sh is identical to what we will be using
- If it does not run on your side, we will not correct your homework
- Note that, if you use any kind of hard-coded paths, this script won't work
- Use paths relative to the project root folder, e.g.:
 - NO: /home/pincopallino/my_folder/model/weights.th
 - OK: model/weights.th









Please be aware that

This is an **individual homework!** Collaboration among the students is **not** allowed.

We will check for **plagiarism** both manually and automatically.

It is **not allowed** to:

- Copy from other students
- Share your code with other students
- Copy from online resources (StackOverflow, GitHub, Medium and so on).

However, you are allowed to use material from **external sources** as long as it is **not central** to the homework.

• In this case, it is **MANDATORY to cite such resources** in the report

Please be aware that

- If we find out that you breached any of the above rules, you will <u>automatically FAIL</u> this homework and you will have to pass a **FULL EXAM**.
- Plagiarism will imply further consequences at the Faculty level.
- While we release the homework on GitHub, **DO NOT FORK THE PROJECT.**
- If you want to continue using GitHub for versioning, clone the project and re-upload it.
- If we realize you shared your code in any way (forking or otherwise), even without the intention of letting others copy, you will be failed automatically.

Use of external data

- You are free to use any additional resources you might need, but they need to be declared in the report
- E.g. XL-WSD or other knowledge resources like SyntagNet



Tips





A few tips to organize your work:

- Start as soon as possible!
 - Training a neural network requires time, possibly hours, depending on your hardware
- Start small!
 - o If you don't get decent results with a very very simple neural network, there is a good chance that adding other things won't make your model perform better
 - Leave the "extras" as the last thing!
- Leave some time for hyperparameter tuning!
 - Sometimes good hyperparameter combinations can do wonders for your neural network
- Use Google Colab (free GPUs!)

Deadline

When to deliver what



Deadline

Submission date: 7 days before the exam date (e.g. for the first session **June 15, 2021** 23:59:59 Italian time (UTC + 1))

Submit the homework through the submission form on Google Classroom. You have to fill the form with the requested information and a link to the zip folder of the homework on Google Drive.





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We will hand out amazing Sapienza NLP t-shirts to the **overall top-5** students!

The final ranking will be computed according to the scores on our **secret** test set.









That's not all

If your work is novel, interesting and original, we will gladly invite you to work together with us to extended on a fully-fledged paper for <u>TOP-TIER</u> <u>INTERNATIONAL CONFERENCE!</u>

Just over the last 12 months, the Sapienza NLP group published more than a dozen of papers!



Questions?

If you have a question that may interest your colleagues, **please ask it on**Google Classroom.

Otherwise, for personal or other questions, send an email to **ALL** of us (but please, only reach for things that can't be asked on the Google Classroom).

Our emails are:

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