

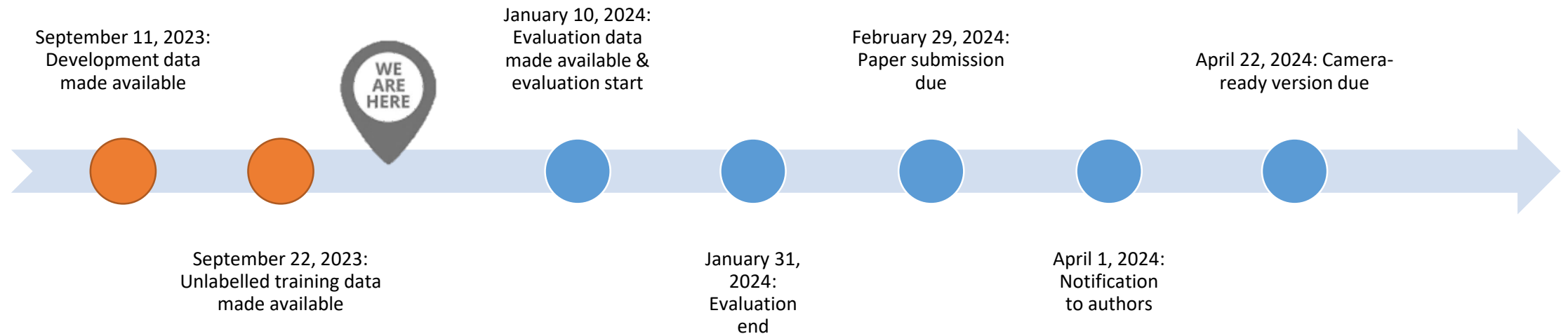
# Hallucination challenge

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# Links de referencia / schedule

- <https://codalab.lisn.upsaclay.fr/competitions/15726>
- <https://semeval.github.io/SemEval2024/>



# GRAN OPORTUNIDAD

SemEval workshop: June 16–21, 2024 (co-located with [NAACL 2024](#) in Mexico City, Mexico).

Esto es lo que pasa cuando  
le dices a DALL-E 3  
“Una conferencia en  
México”



# Enfoque “humano”

- Mi propuesta: que vaya quien más trabaje/mejores resultados obtenga, sea post/predoc, becario...
- Si hay varias líneas interesantes, puede que incluso puedan ir varios subgrupos, uno por paper 😊
- Obviamente, los postdoc tenemos menos tiempo, así que será principalmente “apoyo”: ayudar con la escritura, estado del arte, esbozar las líneas de acción... el cacharreo real (lo diver 😞) será más becarios/predocs, pero obviamente esto cuenta de cara a orden, etc...
- ... porque aquí el que no trabaje no figura. Nada de aparecer el ultimo día, o decir que se es supervisor. Aquí firma el que trabaja.

# Sobre el challenge: evaluación

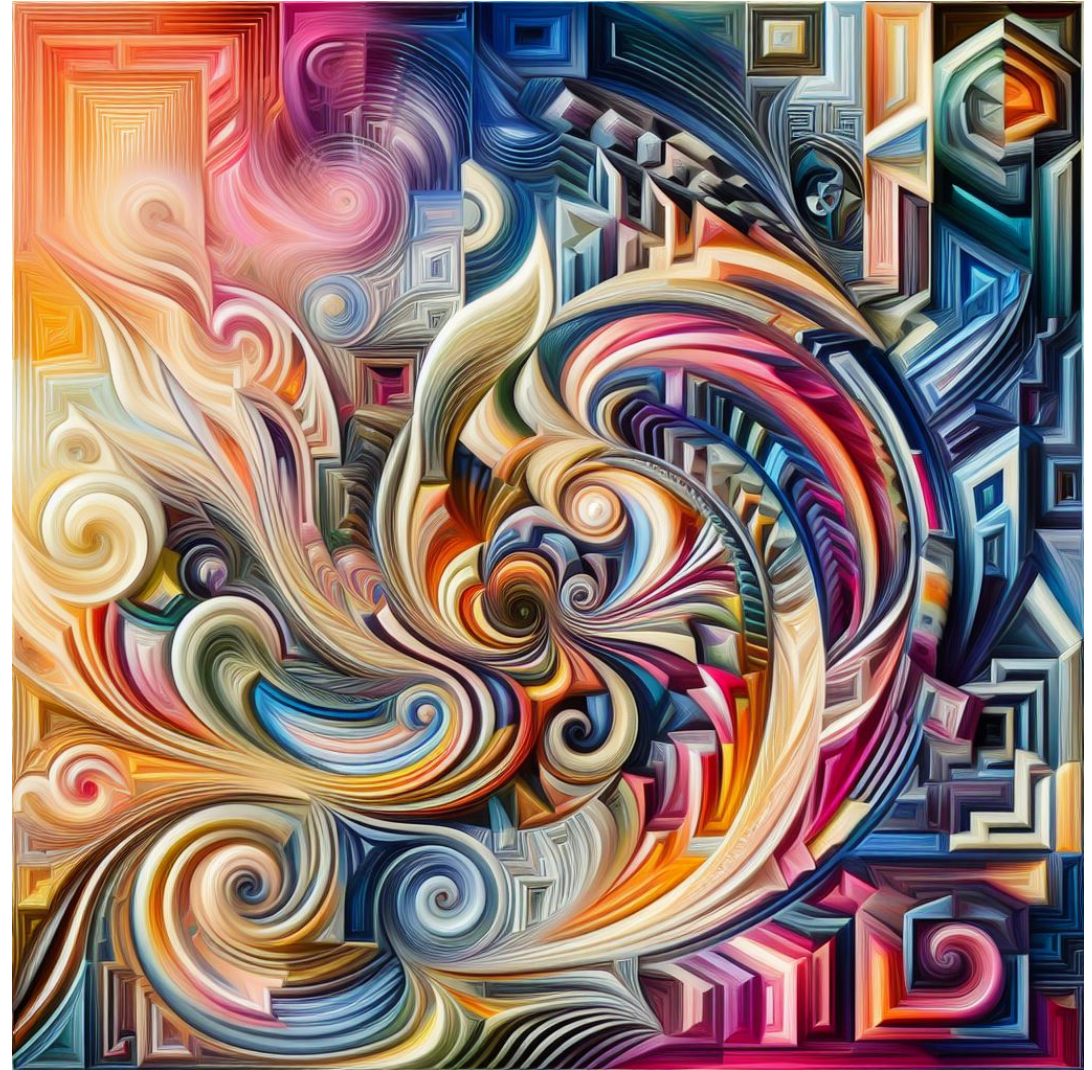
- Submissions will be divided into two tracks:
  - a model-aware track, where we provide a checkpoint to a model publically available on HuggingFace for every datapoint considered
  - a model-agnostic track where we do not. We highly encourage participants to make use of model checkpoints in creative ways.
- For both tracks, all participants' submissions will be evaluated using two criteria:
  - the **accuracy** that the system reached on the binary classification
  - the **Spearman correlation** of the systems' output probabilities with the proportion of the annotators marking the item as overgenerating



# Instructions

Anyone wishing to participate in the task is welcome! Participants will have to:

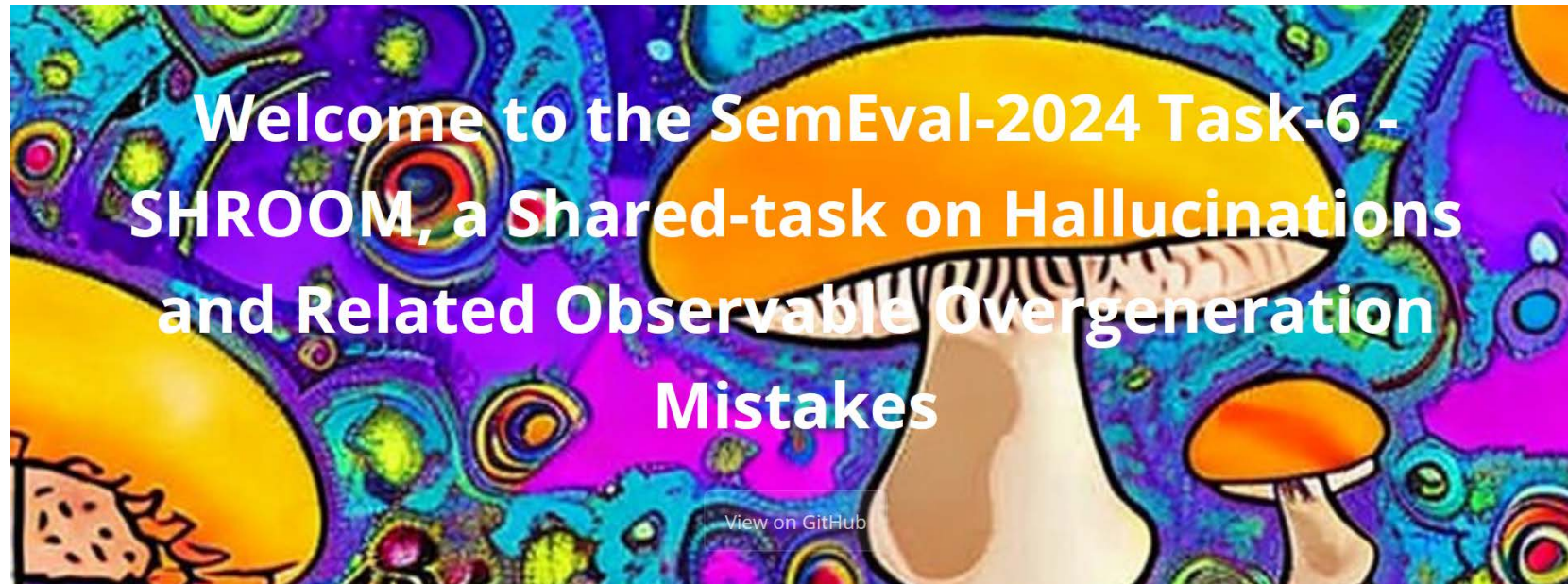
- Submit at least once during the evaluation phase next January;
- Write a system description paper;
- **Review other system description papers (max. 2).**



# About the data

The task consists in a binary classification, where participants are asked to determine whether a given production from an NLP model constitutes a hallucination

<https://helsinki-nlp.github.io/shroom/>



# About the data

- Ya descargado,  
descomprimido y en csv  
(más legible) en:

<https://delicias.dia.fi.upm.es/nextcloud/index.php/s/y9wqgcexPDSDdcM>

```
| JSON2CSV.txt
|
| └── SHROOM_dev-v1
|     README-v1.txt
|     val.model-agnostic.json
|     val.model-aware.json
|
| └── SHROOM_trial-v1.1
|     README-v1.1.txt
|     trial-v1.json
|
| └── SHROOM_unlabeled-training-data-v1
|     train.model-agnostic.json
|     train.model-aware.json
```



# Info available

- a task (`task`), indicating what objective the model was optimized for;
- a source (`src`), the input passed to the models for generation;
- a target (`tgt`), the intended reference "gold" text that the model ought to generate;
- a hypothesis (`hyp`), the actual model production;
- a set of per annotator labels (`labels`), indicating whether each individual annotator thought this datapoint constituted a hallucination or not;
- a majority-based gold-label (`label`), based on the previous per-annotator labels;
- a probability assigned to this datapoint being a hallucination ( $p(\text{Hallucination})$ ), corresponding to the proportion of annotators who considered this specific datapoint to be a hallucination.

# Train dataset (model-aware)

# 30k

Json con array con:

```
{
  "hyp": "Of or pertaining to the official authorities ; governing ; governing ; ",
  "tgt": "Sanctioned by the pharmacopoeia ; appointed to be used in medicine ; officinal .",
  "src": "An official drug or preparation . What is the meaning of official ?",
  "ref": "either",
  "task": "MT",
  "model": "facebook/nllb-200-distilled-600M"
}
```

→ Pred (i.e. a def from Wikipedia)

→ True (i.e. a def from Wikipedia)

→ prompt

NLG tasks:

definition modeling (DM)

machine translation (MT)

paraphrase generation (PG)

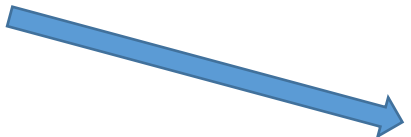
facebook/nllb-200-distilled-600M  
 Itg/flan-t5-definition-en-base  
 tuner007/pegasus\_paraphrase

# Train dataset (model-agnostic)

# 30k

Json con array con:

```
{  
  "hyp": "Don't worry, it's only temporary.",  
  "tgt": "Don't worry. It's only temporary.",  
  "src": "\u041d\u0435  
\u0432\u043e\u043b\u0443\u0439\u0441\u044f. \u0415\u0442\u043e \u0442\u043e\u043b\u044c\u043a\u043e  
\u0432\u0440\u0435\u043c\u0435\u043d\u043d\u043e.",  
  "ref": "either",  
  "task": "MT",  
  "model": ""  
}
```



Either, src or tgt



Siempre ruso en MT!!!  
En el csv se ve  
“Не волнуйся. Это только  
временно.”

# Trial dataset

Json con array con:

```
{
  "hyp": "A district of Kowloon, China."
  "ref": "tgt"
  "src": "The City <define> Chiuchow </define> is Kowloon 's other top restaurant
and is famous for its goose dishes and other specialties from the Chiuchow
region ( you may also wish to try the beef satay done in a creamy sauce ) ."
  "tgt": "The Chaoshan region where the Teochew dialect is spoken."
  "model": ""
  "task": "DM"
  "labels": ["Hallucination", "Hallucination", "Hallucination"]
  "label": "Hallucination"
  "p(Hallucination)": 1.0
}
```

Hallucination  
Not Hallucination

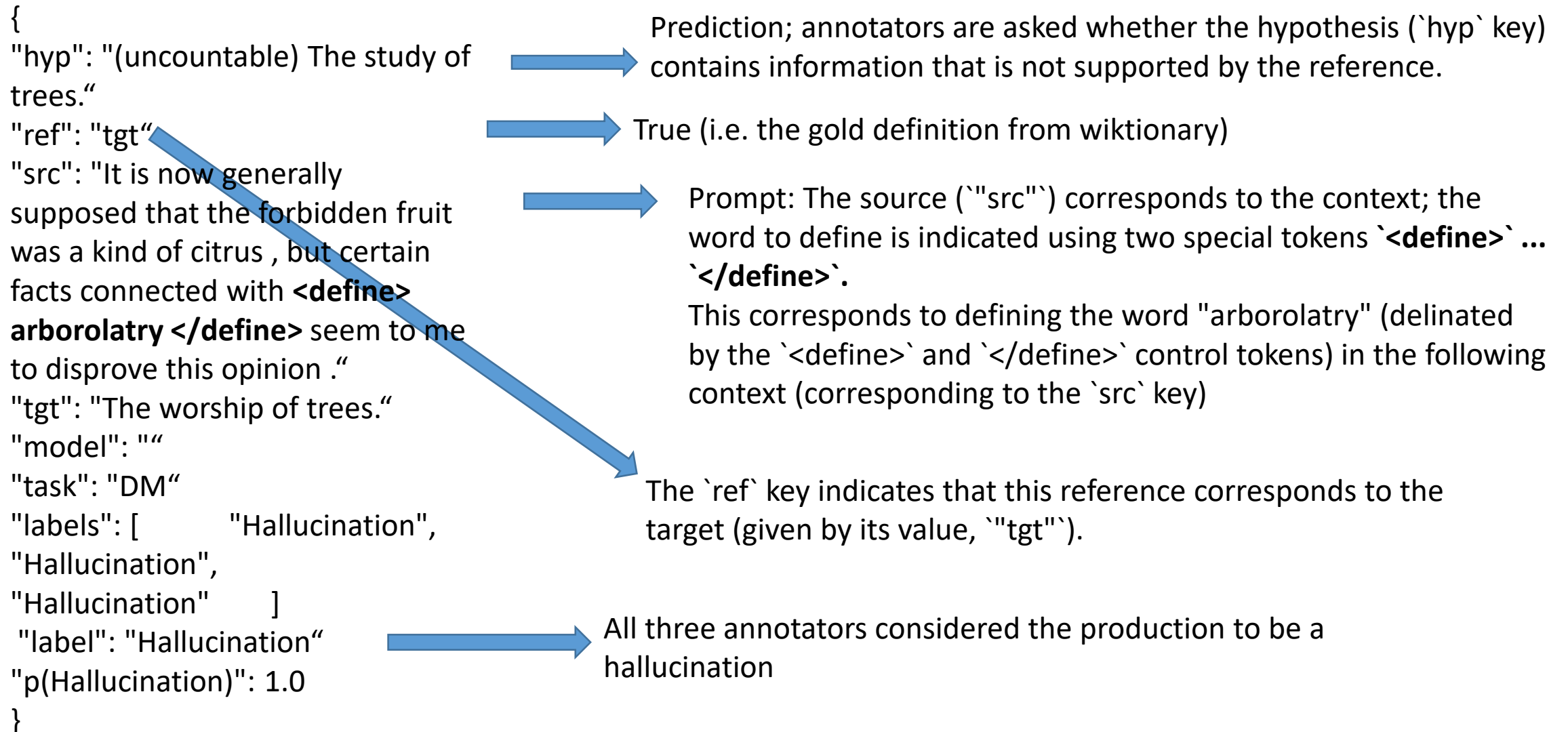
1  
0,66666667  
0,33333333  
0



# Trial readme example (DM)

README TRIAL

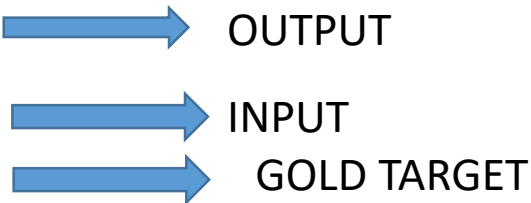
## Definition Modeling



# Trial readme example (PG)

## Paraphrase Generation

```
{  
  "hyp": "When did you see him?",  
  "ref": "either",  
  "src": "When\u2019d you last see him?",  
  "tgt": "When was the last time you saw him?",  
  "model": "tuner007/pegasus_paraphrase",  
  "task": "PG",  
  "labels": [ "Not Hallucination",  
    "Not Hallucination", "Not Hallucination"  
  ],  
  "label": "Not Hallucination",  
  "p(Hallucination)": 0.0  
}
```



For PG datapoints, we also indicate the huggingface model that was used to generate the hypothesis

# Trial readme example (MT)

## Machine Translation

```
{  
  "hyp": "I have nothing to do with it.",  
  "ref": "either",  
  "src": "J'en ai rien \u00e0 secouer.",  
  "tgt": "I don't give a shit about it.",  
  "model": "",  
  "task": "MT",  
  "labels": ["Hallucination", "Not  
Hallucination", "Hallucination"],  
  "label": "Hallucination",  
  "p(Hallucination)": 0.6666666666666666  
}
```

A Traducir: SRC  
Salida: HYP  
TGT es lo correcto

### README TRIAL

The trial set covers datapoints from definition modeling (DM), machine translation (MT) and paraphrase generation (PG). **All other sets should also include text simplification (TS) datapoints.**

Furthermore

- **The train set will not contain manual annotations.**
- The validation and evaluation sets will involve five annotators per datapoint.

# Val folder

- Sólo los tres tipos que conocemos (MT, PG, DM)
- 5 anotadores
- Agnostic: 218 hallucination/281 no hallucination **499**
- Aware: 206 hallucination/295 no hallucination **501**
- Mismos modelos que antes



# Enfoque tecnológico

¿Qué líneas sugerís?

- Distinto enfoque por modelo? (nada garantiza nuevos en test)
- Distinto enfoque por tarea?
- Bases de datos externas? KG, Wikidata, Wikipedia...

¿Organizarse en subgrupos independientes?

- Por tarea
- Por disponibilidad temporal
- Por modelo