# **Named Entity Recognition (NER)**

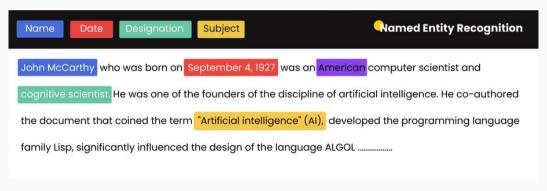
A focus on the GliNER Model Using Twitter Data During COVID-19

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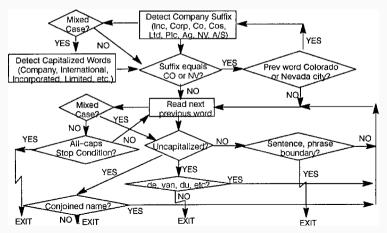
## **Introduction to Named Entity Recognition (NER)**

 Definition: Named Entity Recognition (NER) is an information extraction task (Goyal, Gupta, and Kumar 2018) in Natural Language Processing (NLP) for identifying entities like names of people, locations, organizations, dates, and numerical expressions within text.



#### **Historical overview of NER**

• Early Contributions: Lisa Rau's (1991) work on extracting company names was pioneering, addressing the challenges posed by unknown words in NLP tasks.



## **Historical overview of NER**

- **Origins**: The term "Named Entitty" was formally introduced in the Sixth Message Understanding Conference (MUC-6), where it was recognized as essential for **extracting structured information**, such as names and temporal expressions (Grishman and Sundheim 1996).
- Evolution: Over time, methods evolved from rule-based approaches to machine learning and, eventually, to deep learning-based models (Nadeau and Sekine 2007; Liu, Chen, and Xia 2022).

#### GliNER: a Zero-shot NER Model

 Generalist and Lightweight Model for Named Entity Recognition (Zaratiana et al. 2024)



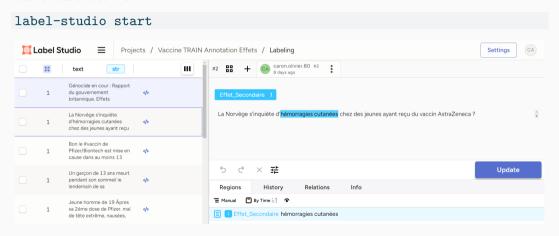
#### **GliNER: Benchmark**

Model	Params	Movie	Restaurant	AI	Literature	Music	Politics	Science	Average
Vicuna-7B	7B	6.0	5.3	12.8	16.1	17.0	20.5	13.0	13.0
Vicuna-13B	13B	0.9	0.4	22.7	22.7	26.6	27.0	22.0	17.5
USM	0.3B	37.7	17.7	28.2	56.0	44.9	36.1	44.0	37.8
ChatGPT	_	5.3	32.8	52.4	39.8	66.6	68.5	67.0	47.5
InstructUIE	11B	63.0	21.0	49.0	47.2	53.2	48.1	49.2	47.2
UniNER-7B	7B	42.4	31.7	53.6	59.3	67.0	60.9	61.1	53.7
UniNER-13B	13B	48.7	36.2	54.2	60.9	64.5	61.4	63.5	55.6
GoLLIE	7B	63.0	43.4	<b>59.1</b>	62.7	67.8	57.2	55.5	58.0
GLiNER-S	50M	46.9	33.3	50.7	60.0	60.9	61.5	55.6	52.7
<b>GLiNER-M</b>	90M	42.9	37.3	51.8	59.7	69.4	68.6	58.1	55.4
GLiNER-L	0.3B	57.2	42.9	57.2	64.4	69.6	72.6	62.6	60.9

Table 1: **Zero-Shot Scores on Out-of-Domain NER Benchmark.** We report the performance of GLiNER with various DeBERTa-v3 (He et al., 2021) model sizes. Results for Vicuna, ChatGPT, and UniNER are from Zhou et al. (2023); USM and InstructUIE are from Wang et al. (2023); and GoLLIE is from Sainz et al. (2023).

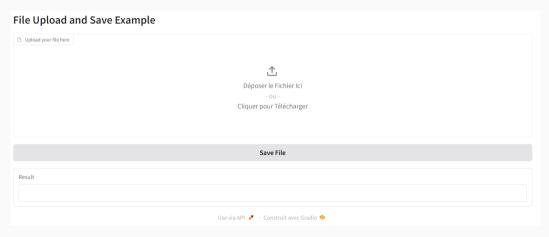
## **Finetuning - Labelling data**

#### Start Label Studio with:

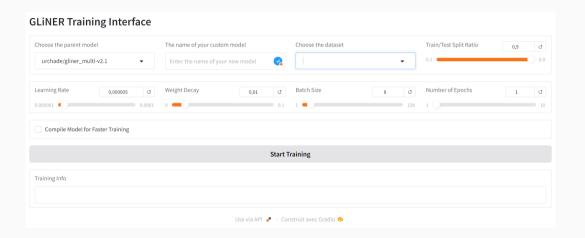


## Finetuning - Upload annotaded data

I used the GliNER Studio Colab notebook (Stepanov and Shtopko 2024) from Knowledgator.



## **Finetuning - Train the model**



## Finetuning - gliner\_multi-v2.1

```
[ ] # Utilisation du modèle préentraîné
    print("\nRésultats avec le modèle préentraîné:")
    pretrained model = load pretrained model()
    # Indiquer le modèle en cours d'utilisation et afficher la configuration
    #print("Modèle en cours: urchade/gliner multi-v2.1")
    #print("Détails de la configuration du modèle préentraîné:")
    #print(pretrained model.config) # Affiche la configuration complète du modèle
    # Exécuter l'annotation et surligner les entités
    entities pretrained = annotate text with char indices(pretrained model, text, labels)
    highlighted text pretrained = highlight entities(text, entities pretrained)
    print(highlighted text pretrained)
    # Afficher les entités détectées
    for entity in entities pretrained:
         print(f"Texte détecté : '{entity['word']}' | Label : {entity['entity']} | Position : {entity['start']}-{entity['end']} | Score : {entity['score']:.2f}")
    Résultats avec le modèle préentraîné:
    Fetching 4 files: 100%
                                                                4/4 [00:00<00:00 144 28it/s]
    Asking to truncate to max length but no maximum length is provided and the model has no predefined maximum length. Default to no truncation.
    Les patients interrogés par Libération témoignaient de nombreux effets secondaires : douleurs abdominales, nausées, vomissements, voire plus de fatigue.
    Texte détecté : 'douleurs abdominales' | Label : Effet Secondaire | Position : 85-105 | Score : 0.89
    Texte détecté : 'nausées' | Label : Effet Secondaire | Position : 107-114 | Score : 0.91
    Texte détecté : 'vomissements' | Label : Effet Secondaire | Position : 116-128 | Score : 0.88
```

## Finetuning - Custom model for Covid-19 vaccines' side effects

```
[9] # Utilisation du modèle personnalisé
        print("Résultats avec le modèle personnalisé:")
        custom model path = "drive/MyDrive/Models/custom model"
        custom model = load custom model(custom model path)
        # Indiquer le modèle en cours d'utilisation et afficher la configuration
        #print(f"Modèle en cours: {custom model path}")
        #print("Détails de la configuration du modèle personnalisé:")
        #print(custom_model.config) # Affiche la configuration complète du modèle
        # Exécuter l'annotation et surligner les entités
        entities custom = annotate text with char indices(custom model, text, labels)
        highlighted text custom = highlight entities(text, entities custom)
        print(highlighted text custom)
        # Afficher les entités détectées
        for entity in entities custom:
            print(f"Texte détecté : '{entity['word']}' | Label : {entity['entity']} | Position : {entity['start']}-{entity['end']} | Score : {entity['score']:.2f}")
   → Résultats avec le modèle personnalisé:
        config.ison not found in /content/drive/MyDrive/Models/custom model
        WARNING: huggingface hub.hub mixin: config. ison not found in /content/drive/MyDrive/Models/custom model
        Asking to truncate to max length but no maximum length is provided and the model has no predefined maximum length. Default to no truncation.
        Les patients interrogés par Libération témoignaient de nombreux effets secondaires : douleurs abdominales, nausées, vomissements, voire plus de fatigue.
        Texte détecté : 'douleurs abdominales' | Label : Effet Secondaire | Position : 85-105 | Score : 0.97
        Texte détecté : 'nausées' | Label : Effet Secondaire | Position : 107-114 | Score : 0.97
        Texte détecté : 'vomissements' | Label : Effet Secondaire | Position : 116-128 | Score : 0.98
        Texte détecté : 'fatigue' | Label : Effet Secondaire | Position : 144-151 | Score : 0.83
```

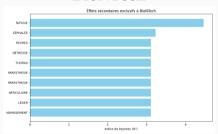
### **Pharmaceutical Brands**

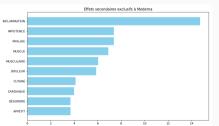
# Pfizer AstraZeneca Moderna Sanofi BioNTech FATIGUE FATIGUE CEPHALE COURBATURE ALERGY CONVENTION SOUL CONNECTION FACIAL FOR ANALYSIE COURBATURE ALERGY CONVENTION FACIAL FOR ANALYSIE OUL EUR MAL CEPHALETETE FEVER ASTRACTOR FACIAL BIONTECH COURBATURE COURBATURE FIVER ETE MAL FOR ANALYSIE FACIAL FOR ANALYSIE FOR ANAL

# **Pharmaceutical Companies**

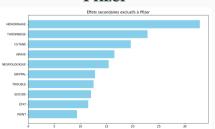


#### **BioNTech**





#### **Pfizer**



## StreamLit App for your data

- $\bullet\ https://hugging face.co/spaces/olivier caron/GLiNER\_file$
- Clone the repository on your own computer:
- https://github.com/oliviercaron/GliNER\_streamlit

# Bibliography

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