Dense retrieval system for general court laws

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Overview

- Baseline system simple retrieval of k most similar contexts
- Dense retrieval model retrieval + re-ranking of the outputs based on the synthetic nli dataset generated from our questions and contexts
- Creation of a first draft of the front-end
- Automatic and Human Evaluation

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Front-end

- Used streamlit to create front-end
- Users can select any one of the 3 models and get related contexts
- Users are able to select the number of contexts they want to retrieve
- Online version hosted on HuggingFace spaces, running on CPU

Front-end



Context: la convocation a une assemblee generale de coproprietaires concernant des epoux proprietaires d'un lot dependant de leur communaute de biens doit etre libellee au nom des deux

Context: le juge n'est pas tenu , en cas de desaccord des parents , d'ordonner la residence en alternance a titre provisoire

Context: l'indemnite d'eviction due au locataire doit etre evaluee en excluant la taxe sur la valeur ajoutee du chiffre d'affaires qui a servi de base a son calcul si tels sont les usages de la profession exercee par le preneur

Context: pour le calcul de la surface maximale prevue par l'article l. 12 - 6 du code rural et de la peche

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Human Evaluation

- Human evaluation done using likert scale (0-5)
- Same interface as front-end but with added rating feature
- Users rate the retrieved information of each model, but are unaware of model type
- User rating data is stored in a json file
- Storing query, retrieved contexts, model type and rating
- No personal information is stored

Automatic Evaluation

- Using embedding based scoring system, similar to BERTScore[2], DepthScore[1]
- Compute similarity of query and contexts as a average of cosine similarity between their embeddings
- Use different model sentence transformer model¹ unseen by our system to get embeddings
- Compute average score from a list of queries to get overall score
- Compare results among the three models
- Preliminary result based on 14 questions

Models	Mean Similarity Score	Std.
Civile-law-IR	0.88	0.018
STSB	0.83	0.022
DR-Baseline	0.86	0.019

Table: Overall Mean similarity score and Standard deviation for 3 dense retrieval models

¹https://huggingface.co/nreimers/MiniLM-L6-H384-uncased

Automatic Evaluation results

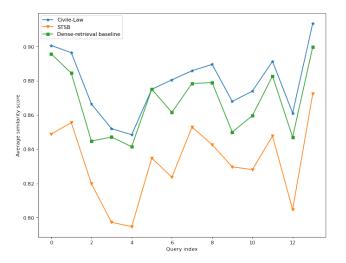


Figure: Comparison based on mean similarity score for the 3 dense retrieval models

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Timeline

- Jan 23 Jan 27:
 - ► Complete human evaluation
 - ► Prepare the defense
 - Final report corrections

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References



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Thank you for your attention !