



Chère maison or maison chère?

Transformer-based Prediction of Adjective Placement in French

Eleni Metheniti, Tim Van de Cruys, Wissam Kerkri,
Juliette Thuilier, Nabil Hathout

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Position of attributive adjectives in French

ma chère maison = 🏠❤️ vs. *ma maison chère* = 🏠💰

- (Mostly) anteposed adjectives:
 - Ordinals: *troisième* 'third'
 - High-frequency adjectives: *grand* 'big'
- (Mostly) postposed adjectives:
 - Colors: *rouge* 'red'
 - Polysyllabic adjectives: *fabuleux* 'fabulous'
- Mobile adjectives (w.r.t. semantics):
 - e.g. *ancien* 'old, former' or 'ancient'
 - e.g. *cher* 'dear' or 'expensive'

Abeillé and Godard (1999); Thuilier (2013); Benzitoun (2014)

Our research questions

- Are transformer-based embeddings sensitive to word order, when positional information is semantically important?
- Are they already or do they need to learn additionally?
- How do their decisions compare to native speakers?

Exp. 1: Finetuning & Classification of adjective position

- Input: Two sentences with different word order in noun-adjective pair (labels: 0 = anteposition, 1 = postposition)

*On construit les éléments de **haut niveau**.* - 0 - (original)

*On construit les éléments de **niveau haut**.* - 1 - (permuted)

- Also finetune with one-sentence input (original)
- Also finetune with attention masks:
 - Mask entire context except for noun and adjective
 - Mask noun and adjective, context visible
- Baselines: Logistic regression, CNN, frequency

French Transformer-based models

- CamemBERT (Martin et al., 2020): RoBERTa-based model
 - camembert-base
 - camembert-large
- FlauBERT (Le et al., 2020): BERT-based, XLM elements
 - flaubert-small-cased
 - flaubert-base-cased
 - flaubert-base-uncased
 - flaubert-large-cased

Datasets

- frWaC (Baroni et al., 2009)
- Universal Dependencies 3.0 (Zeman et al., 2021)
- Different training and test sets for finetuning:

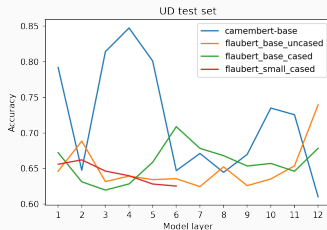
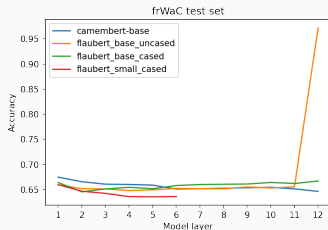
Source	Train	Val.	frWaC test	UD test (100%)	UD test (25%)
frWaC	76,164	7,672	7,740	19,437	5,151
frWaC +UD (75%)	91,615	7,672	7,740	-	5,151
UD (75%)	13,905	1,546	7,740	-	5,151

Exp. 1: Finetuning & Classification of adjective position

- **Results:**
 - 0.87-0.99 on frWaC
 - 0.97-0.99 on frWaC+UD
 - 0.62-0.99 on UD
 - CamemBERT > FlauBERT, but baselines close
- Masking context only good for CamemBERT, catastrophic for FlauBERT
- Masking noun-adjective had overall lower results, but not failure
- Error analysis: few mistakes, some in mobile adjectives, some from parsing

Exp. 2: Testing adjective pretrained embeddings

- **Classification only with adjective embeddings** (and logistic regression): moderate, successful only circumstantially



Exp. 2: Testing adjective pretrained embeddings

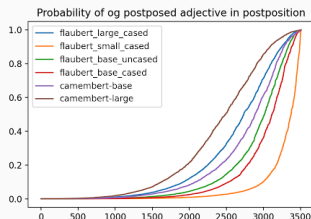
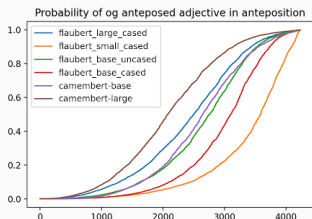
- MLM adjective probabilities:

Probability of *haut* in the masked position:

On construit les éléments de [MASK] niveau. (original)

*On construit les éléments de **niveau** [MASK].* (permuted)

- Higher probability of masked adjective in original position than opposite, postposed adjectives preferred postposition more (note correlation of anteposition & frequency)



Exp. 3: Human judgments vs models' probabilities

- Same two-sentence setup as Experiment 1
- Dataset of challenging/control sentences:
 1. Adjective/Noun dependents
 2. Fixed expressions
 3. Structural persistence
 4. Blocked/mobile adjectives
- 3 questionnaires, a total of 71 human participants

Exp. 3: Human judgments vs models' probabilities

Correlation of human judgments and model classification probabilities:

Model	1	2	3	4
camembert-base	0.21	-0.19	-0.08	0.47
camembert-large	0.67	0.61	0.53	0.51
flaubert_small_cased	0.51	-0.03	0.16	0.78
flaubert_base_cased	0.52	0.09	0.38	0.71
flaubert_base_uncased	0.40	0.22	0.63	0.56
flaubert_large_cased	0.46	0.18	0.63	0.47

Discussion

- Easy task... until it's not! Frequency is key
- Finetuning: data-hungry, multiple domains a plus
- Context is crucial and is exploited by models
- But not enough information in the adjective embedding
- Models vs. Humans:
 - Unacceptable mistakes
 - Models have too high probabilities
 - Models prefer postposition, even when wrong

Thank you for your (unmasked) attention!

Code for this paper:

https://github.com/lenakmeth/word_order

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