

Endangered Languages—How much data do we need to model them well?

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Introduction

- More than 43% of languages spoken in the world are endangered (Zhang et al., 2022)
- What if we could use NLP to preserve these languages?
 - Difficult to train a model from scratch on minimal data
 - What about using a pre-trained model in a similar language?
- Can we find the minimum amount of tokens required for a pre-trained model to perform well in another language?
- Use a pre-trained English model, fine-tune it on French data

What is considered an endangered language?

- Open Super-large Crawled Aggregated coRpus (OSCAR)
- 153 languages
- 13% are considered vulnerable or endangered

Language Endangerment Level	Average Number of Tokens	Standard Deviation of Tokens
Not endangered	8.130 billion	46.938 billion
Vulnerable	13.878 million	48.027 million
Definitely endangered	28.353 million	54.083 million
Severely endangered	949 thousand	941 thousand
Critically endangered	6,347	-

Christopher Moseley. Atlas of the world's languages in danger. UNESCO. 2010.

Main Resources

- Source language: English
- RoBERTa
 - “roberta-base”
 - Monolingual and not fine tuned
- SQuAD
 - Stanford Question Answering Dataset

- Target language: French
- CamemBERT
 - “illuin/camembert-base-fquad”
 - Use to benchmark a good performance
- FQuAD
 - French SQuAD equivalent

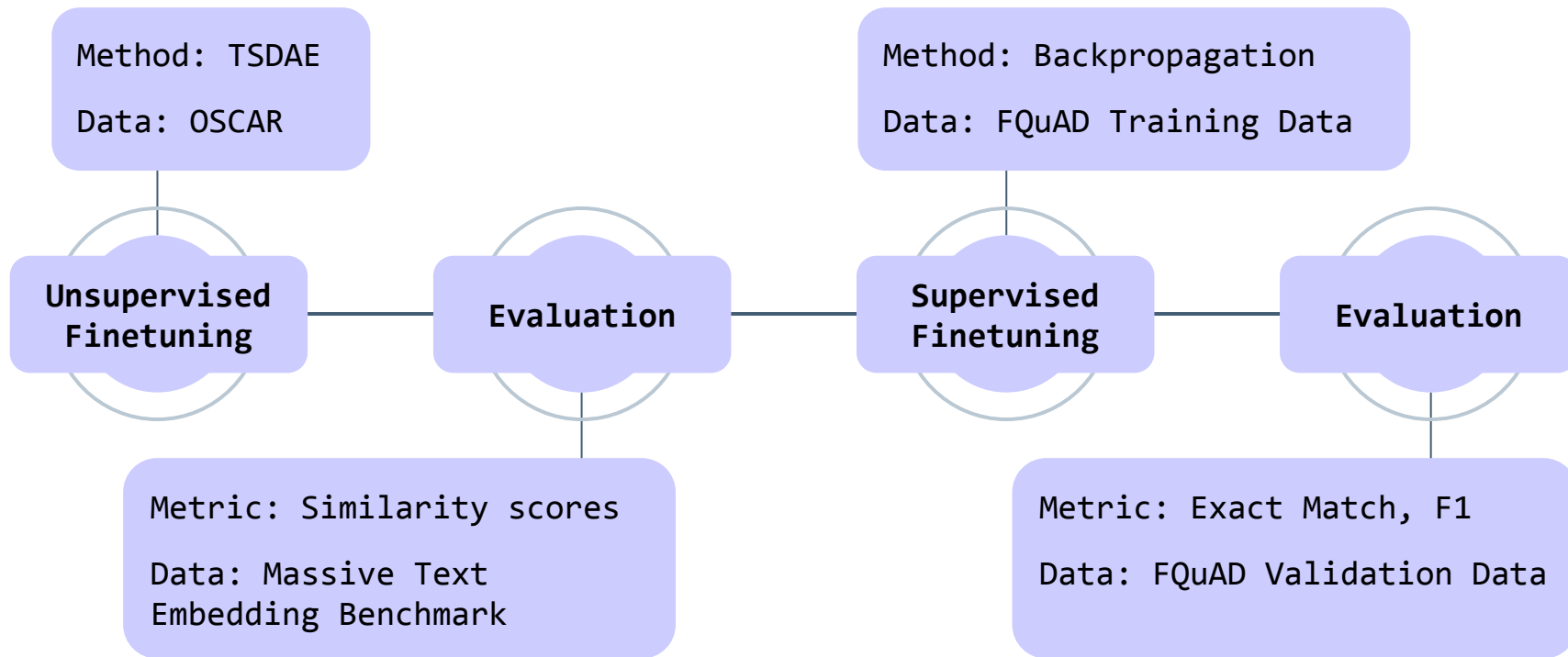
The literal meaning of Durbar Square is a "place of palaces". There are three preserved Durbar Squares in Kathmandu valley and one unpreserved in Kirtipur. The Durbar Square of Kathmandu is located in the old city and has heritage buildings representing four kingdoms (Kantipur, Lalitpur, Bhaktapur, Kirtipur); the earliest is the Licchavi dynasty. The complex has 50 temples and is distributed in two quadrangles of the Durbar Square. The outer quadrangle has the Kasthamandap, Kumari Ghar, and Shiva-Parvati Temple; the inner quadrangle has the Hanuman Dhoka palace. The squares were severely damaged in the April 2015 Nepal earthquake.

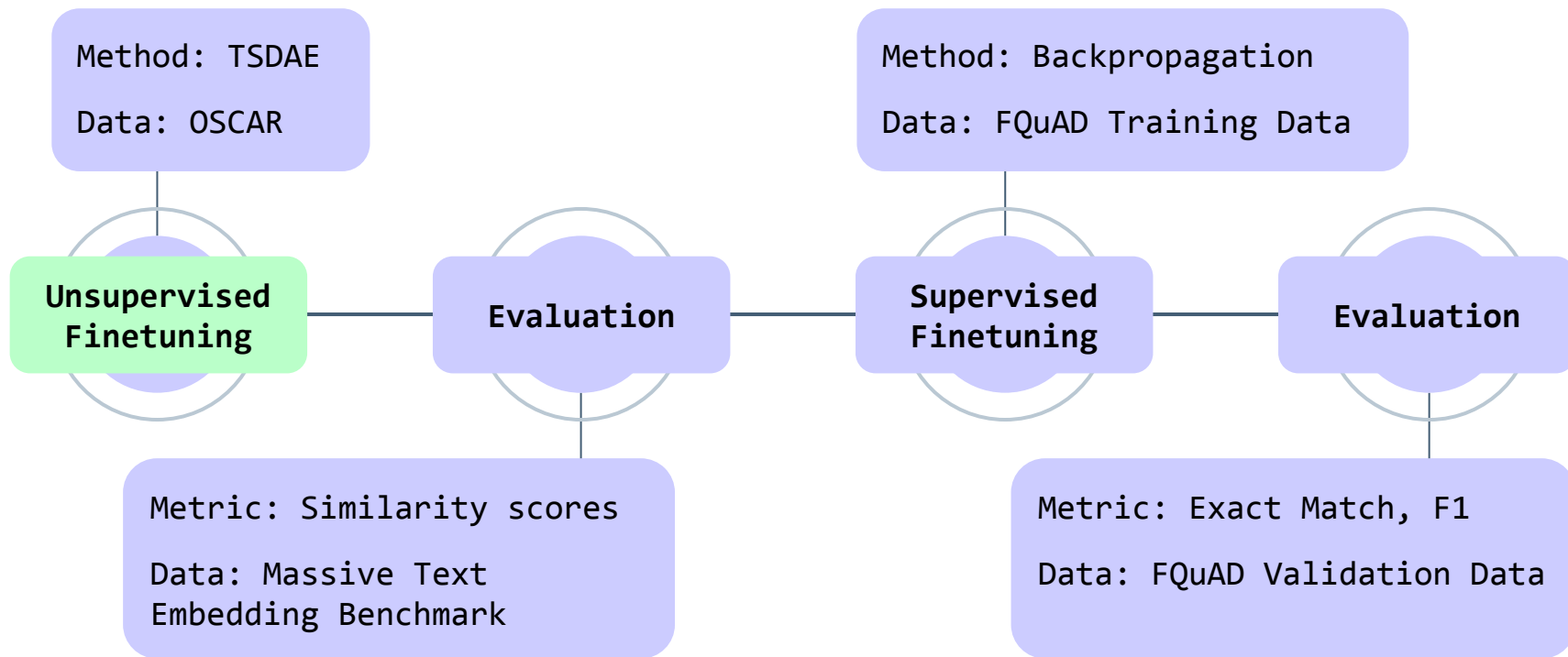
When did a notable earthquake occur that damaged Kathmandu's Durbar Square?

Les deux tableaux sont certes décrits par des documents contemporains à leur création mais ceux-ci ne le font qu'indirectement car ils concernent principalement La Vierge aux rochers. Aussi demeurent-ils objets de spéculations pour les chercheurs quant à leur statut de première ou seconde version de l'œuvre, leur création, leur attribution, leur datation, leur disposition exacte sur le retable et les raisons qui ont poussé à leurs modifications au cours du temps – notamment pour ce qui concerne la couleur du fond.

Que concerne principalement les documents?

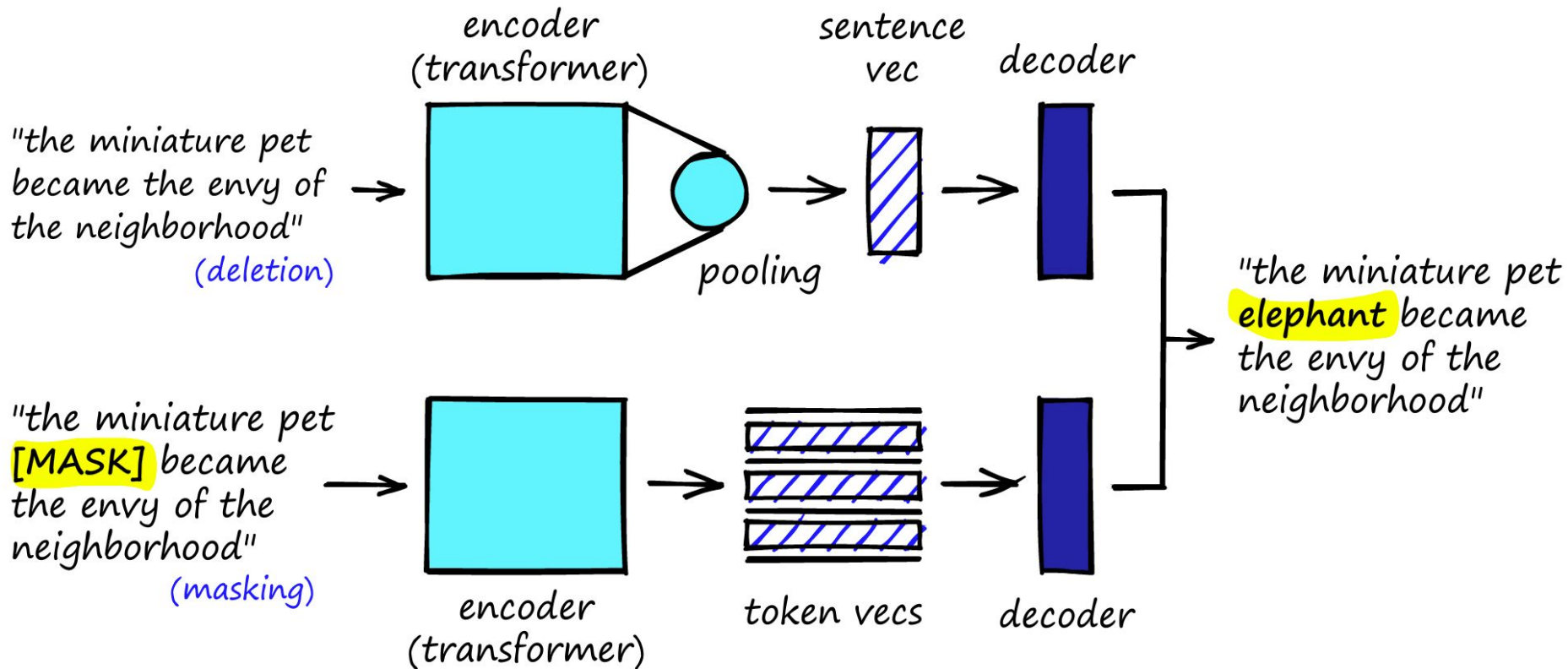
Methodology





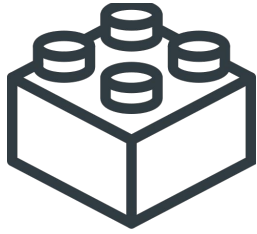
OPTIONS BINAIRES EN LIGNE. EBOOK. Cher trader, Bienvenue sur Options Binaires en ligne, ceci est le premier volet d'une série de 5 Ebooks pour apprendre à négocier des options binaires en ligne, nous sommes heureux de vous introduire dans le monde du négoce financier grâce aux options binaires. Les options binaires sont connues comme étant un moyen super rapide, simple et accessible pour investir et gagner de l'argent en ligne. En effet, vous pouvez vous lancer dans les options binaires, et ce, même si vous êtes nul en trading. Voyez comment il est possible d'investir facilement dans les options binaires. Nous allons y voir ce qu'est une option binaire, les pièges à éviter pour un investissement dans l'option binaire, les différentes options binaires ainsi que toutes les astuces pour maximiser vos chances de réussir votre investissement. En même temps, nous allons également vous donner de bons conseils sur le choix des plateformes d ... Avantages à investir dans les options binaires. Le premier avantage des options binaires réside dans leur simplicité : il suffit d'estimer la direction qu'une option va prendre. Sur les actions traditionnelles, on spéculé sur une différence de prix réel, beaucoup plus difficile à prédire. Investir dans les options binaires : exemple d'option binaire À titre d'exemple, nous pourrions considérer une option binaire associée à l'or. La valeur monétaire de cet or est estimée à 1300 dollars et est associée à une option qui s'élève à 100 dollars.

Excerpt from Open Super-large Crawled Aggregated coRpus (OSCAR)

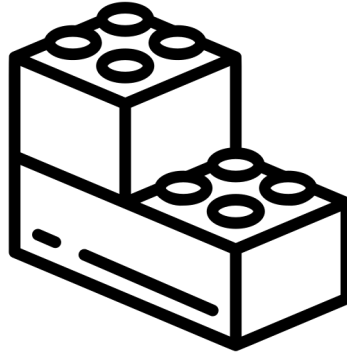


Transformer(-based) and Sequential Denoising Auto-Encoder (TSDAE)

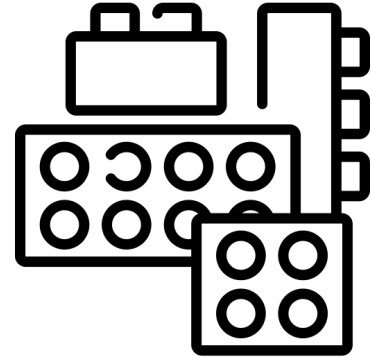
What roberta-base was finetuned on



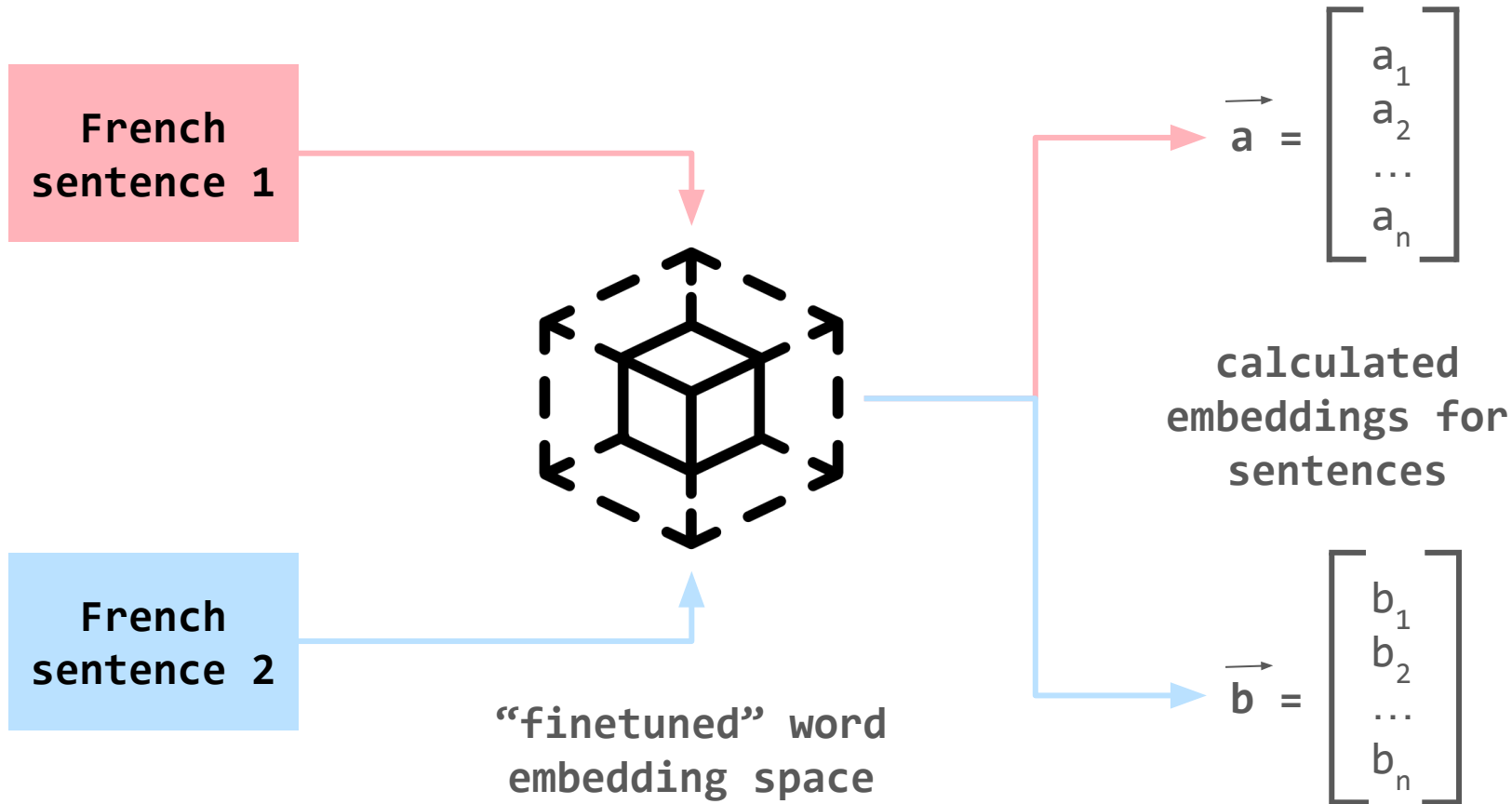
**6,500
tokens**



**100,000
tokens**



**950,000
tokens**



**cosine
distance**

how different
embeddings are
from one
another

$$= 1 - \cos(\vec{a}, \vec{b})$$

how similar
embeddings are
to another

Shkhanukova, Milana. "Cosine distance and cosine similarity."

<https://medium.com/@milana.shkhanukova15/cosine-distance-and-cosine-similarity-a5da0e4d9ded>

The Dangers of a High Learning Rate

Number of Tokens Finetuned on	Correlation Score
Control: roberta-base	0.382
6,500_3e-5_1epoch	0.376
6,500_3e-7_1epoch	0.398
6,500_3e-7_2epoch	0.398
100,000_3e-5_1epoch	0.349
100,000_3e-7_2epoch	0.418
100,000_3e-7_3epoch	0.406
950,000_3e-5_1epoch	0.124
950,000_3e-7_1epoch	0.405
950,000_3e-10_1epoch	0.398
950,000_3e-7_2epoch	0.194
Camembert	0.634

Table.1: Evaluation of Unsupervised Finetuning

*Model naming convention: token amount, learning rate, followed by epoch amount

More Tokens, More Complex Hyperparameters

Lowering learning rate and increasing epochs improve performance	Number of Tokens Finetuned on	Correlation Score	No marked difference
	Control: roberta-base	0.382	
	6,500_3e-5_1epoch	0.376	
	6,500_3e-7_1epoch	0.398	
	6,500_3e-7_2epoch	0.398	Increasing epochs decreases performance
	100,000_3e-5_1epoch	0.349	
	100,000_3e-7_2epoch	0.418	
	100,000_3e-7_3epoch	0.406	
	950,000_3e-5_1epoch	0.124	
	950,000_3e-7_1epoch	0.405	
	950,000_3e-10_1epoch	0.398	
	950,000_3e-7_2epoch	0.194	
Camembert		0.634	

Table.1: Evaluation of Unsupervised Finetuning

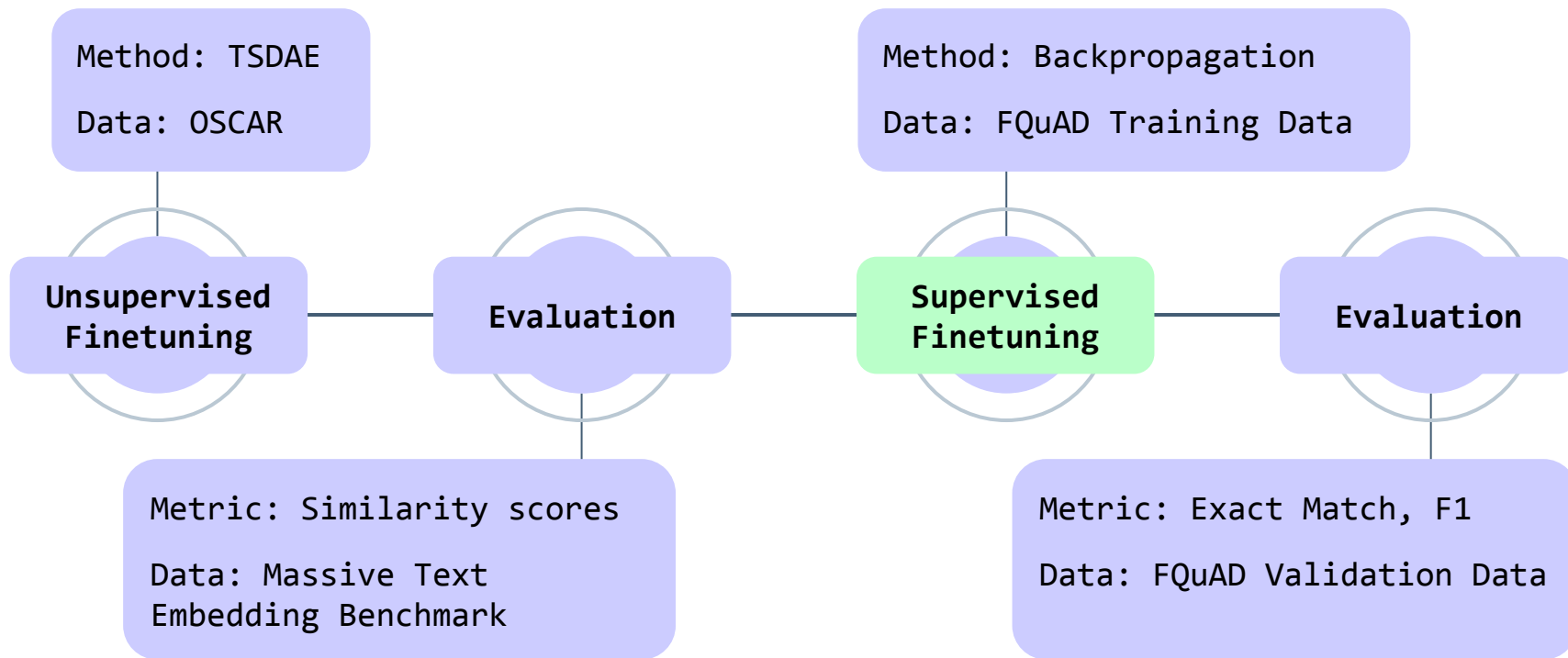
*Model naming convention: token amount, learning rate, followed by epoch amount

Word Embeddings did not improve significantly

Number of Tokens Finetuned on	Correlation Score
Control: roberta-base	0.382
6,500_3e-5_1epoch	0.376
6,500_3e-7_1epoch	0.398
6,500_3e-7_2epoch	0.398
100,000_3e-5_1epoch	0.349
100,000_3e-7_2epoch	0.418
100,000_3e-7_3epoch	0.406
950,000_3e-5_1epoch	0.124
950,000_3e-7_1epoch	0.405
950,000_3e-10_1epoch	0.398
950,000_3e-7_2epoch	0.194
Camembert	0.634

Table.1: Evaluation of Unsupervised Finetuning

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Context

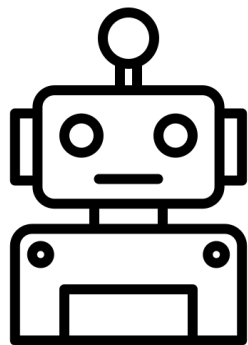
Piazzi observa Cérès 24 fois, la dernière fois le 11 février. Le 24 janvier 1801, Piazzi annonça sa découverte par des lettres à plusieurs collègues italiens, parmi lesquels Barnaba Oriani à Milan. Il la décrivit comme une comète, mais remarqua que « puisque son mouvement est lent et uniforme, il m'a semblé à plusieurs reprises qu'il pourrait s'agir de quelque chose de mieux qu'une comète. » En avril, Piazzi envoya ses observations complètes à Oriani, Bode et Lalande à Paris. Elles furent publiées dans l'édition de septembre 1801 du *Monatliche Correspondenz*.

Question

Pourquoi Cérès n'était pas directement assimilable à une comète ?

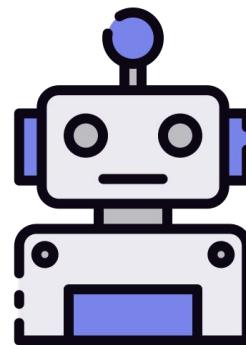
Answer

son mouvement est lent et uniforme



roberta-base

- Clean copy of roberta-base
- Never undergone unsupervised finetuning
- Never undergone supervised finetuning



roberta-base-fquad-finetuned

- **Modified** copy of roberta-base
- Never undergone unsupervised finetuning
- **Has undergone supervised finetuning**

Supervised finetuning improves score the most

Our method of
improving the word
embeddings does
not have much
impact

Model	FQUAD Exact Match	FQUAD F1
Control: roberta-base	0.063%	7.58%
roberta-base-fquad-finetuned	21.6%	31.9%
6,500_3e-5_1epoch	21.1%	31.9%
6,500_3e-7_1epoch	22.2%	32.4%
6,500_3e-7_2epoch	21.8%	31.8%
100,000_3e-5_1epoch	21.4%	31.8%
100,000_3e-7_2epoch	21.0%	31.5%
100,000_3e-7_3epoch	21.7%	32.9%
950,000_3e-10_1epoch	21.6%	32.2%
950,000_3e-7_1epoch	21.5%	32.1%
950,000_3e-7_2epoch	21.4%	32.1%
Camembert	45.8%	68.2%

Table.2: Evaluation of Supervised Finetuning

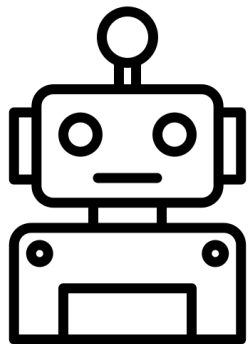
*Model naming convention: token amount, learning rate, followed by epoch amount

Varying token size does not improve performance

Model	FQUAD Exact Match	FQUAD F1
Control: roberta-base	0.063%	7.58%
roberta-base-fquad-finetuned	21.6%	31.9%
6,500_3e-5_1epoch	21.1%	31.9%
6,500_3e-7_1epoch	22.2%	32.4%
6,500_3e-7_2epoch	21.8%	31.8%
100,000_3e-5_1epoch	21.4%	31.8%
100,000_3e-7_2epoch	21.0%	31.5%
100,000_3e-7_3epoch	21.7%	32.9%
950,000_3e-10_1epoch	21.6%	32.2%
950,000_3e-7_1epoch	21.5%	32.1%
950,000_3e-7_2epoch	21.4%	32.1%
Camembert	45.8%	68.2%

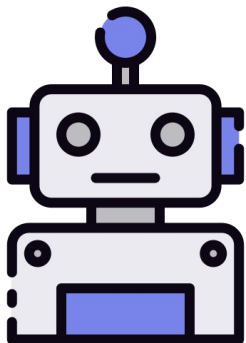
Table.2: Evaluation of Supervised Finetuning

*Model naming convention: token amount, learning rate, followed by epoch amount



roberta-base
responses to
first 11 FQuAD
QA pairs

1. mais ceux-ci ne le font qu'indirectement car ils concernent principalement La Vierge aux
2. mais ceux-ci ne le font qu'indirectement
3. mais ceux-ci ne le font qu'indirectement car ils concernent principalement La Vierge aux
4. empty
5. ans
6. empty
7. dans la version
8. dans la version londonienne du panneau
9. dans la version londonienne du panneau
10. dans la version
11. puisque ce dernier fait partie des trois artistes désignés dans le contrat de commande, chacun ayant un rôle

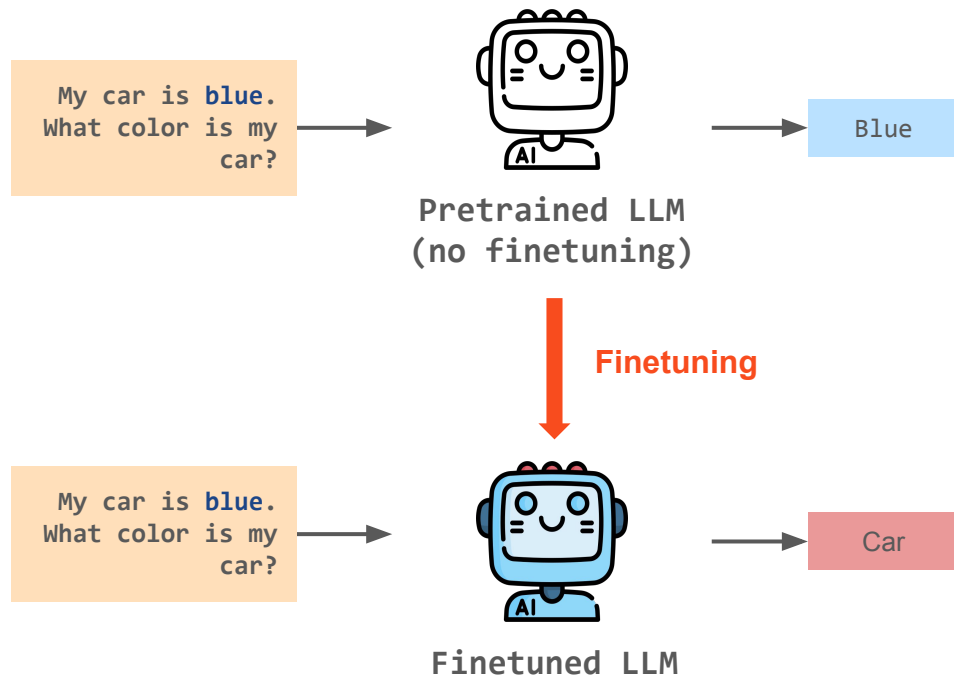


roberta-base-fquad
-finetuned
responses to first 11
FQuAD QA pairs

1. La Vierge aux rochers
2. documents contemporains à leur création
3. objets de spéculations
4. droite
5. gauche
- 6.
7. l'atelier de Léonard de Vinci
- 8.
- 9.
10. (La Vierge aux rochers)
11. trois

Catastrophic Forgetting

Catastrophic Forgetting



When finetuning causes models to forget what they learned during pretraining

Results of Catastrophic Forgetting

Model	SQUAD Exact Match	SQUAD F1
Control: roberta-base	0.194%	4.33%
roberta-base-fquad-finetuned	42.0%	46.1%
6,500_3e-5_1epoch	42.6%↑	46.5%↑
6,500_3e-7_1epoch	45.0%↑	48.6%↑
6,500_3e-7_2epoch	43.8%↑	47.7%↑
100,000_3e-5_1epoch	43.0%↑	47.0%↑
100,000_3e-7_2epoch	39.4%↓	43.9%↓
100,000_3e-7_3epoch	41.7%↓	45.7%↓
950,000_3e-10_1epoch	41.8%↓	45.9%↓
950,000_3e-7_1epoch	42.5%↑	46.4%↑
950,000_3e-7_2epoch	44.0%↑	47.8%↑
roberta-base-squad2	79.5%	82.5%

Table.3: Catastrophic Forgetting

*Model naming convention: token amount, learning rate, followed by epoch amount

Minor catastrophic forgetting:

- Most models did not decrease in performance
- All performance decreases were within 3 percentage points of roberta-base-fquad-finetuned

Ideas for Next Steps

Supervised
finetuning on more
epochs

Unsupervised
finetuning on more
tokens

Manual evaluation
of outputs;
qualitative analysis

What if we had a base model trained in an endangered language?

If we had ample text in a related language...

Finetune the base model on the text of the related language

Would language similarities improve model performance on the original language?

Issues:

We would need an evaluation set for the original language (which is already low-resource)

Ethical Implications

- Our current work is theoretical
- Essential to consider whether communities who use low-resource/endangered languages actually **want** technology made for their language (Wilson, 2022)
- Language is far more than just “data”

Thank you!

Any questions?

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