

Fall Detection in EHR using
Word Embeddings and Deep Learning

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Brief context ...



Multidisciplinary Group and ...

PUCRS

Pontifical Catholic University - Rio Grande do Sul

PPGCC and PPGGB

Graduate program in Computer Science Graduate program in Biomedical Gerontology



... and Multicentric Group

















Scope of this Research

Fall

This adverse event is very common in the hospital environment. The Morse Fall Scale measure the risk of patient fall.

There is a underreport number of fall event notifications.



Report System





Underreported Events

Events are notified voluntarily in the system

Source of useful information to improve care and subsidize continuing education!

Due to lack of knowledge, forgetfulness or lack of time, it is underreported. Only 10-20% of these events are reported.



IHI Global Trigger Tool

Institute for Healthcare Improvement

"This tool includes a list of known AE triggers as well as **instructions for selecting records**, training information, and appendices with references and common questions."



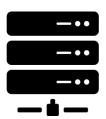
Objectives



Generate
algorithms and
models for fall
events
detection



Make available fall events datasets for shared tasks



Evaluate models within Hospitals

Experiments ...



Annotation Process - WebAnno



^{*} not a real fall report example



Annotation Process - Results





Annotation Process - Results

Fall per Patient in Annotated Dataset

# of Patients	% of Total	# of Falls
316	87.0 %	1 fall
36	10.0%	2 falls
11	3.0 %	3 falls
1	0.3 %	4 falls
•••	•••	•••



Fall Detection - Experiments

Annotation of 1,078 Clinical Notes

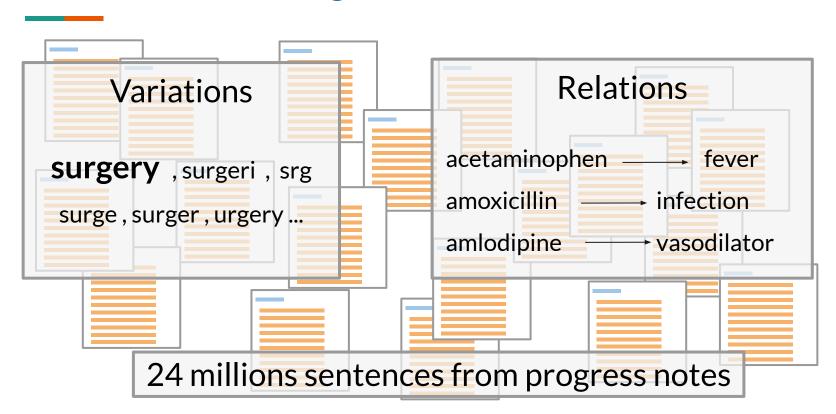
Evaluation of three Word Embedding Models

- Wikipedia-PT
- NILC (Wikipedia + 20 sources)
- GHC: 23 millions sentences from Clinical Notes

State-of-the-art NLP Neural Network (LSTM)

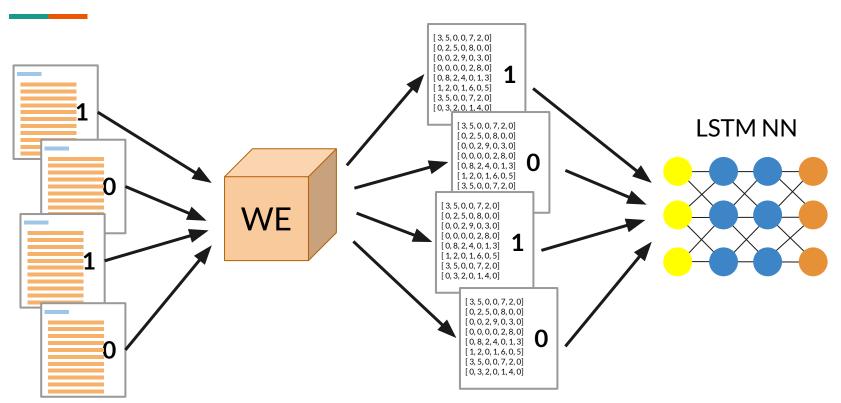


Word Embeddings



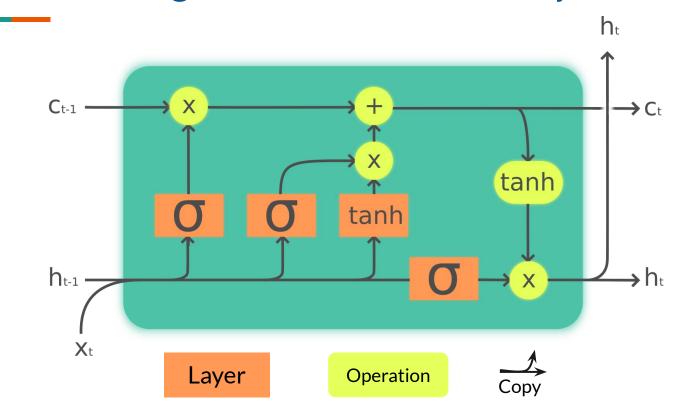


Word Embeddings Encoding





LSTM (Long Short-Term Memory)



Results ...



Fall Detection - Results

Embeddings 5-Fold Cross Validation: F-Measure

	Word2Vec	FastText
Wikipedia	0.88 ± 0.14	0.87 ± 0.11
NILC (broad-domain)	0.77 ± 0.06	0.89 ± 0.13
EHR (specific-domain)	0.88 ± 0.14	0.90 ± 0.13
Baseline (RandomForest)	0.73 ± 0.03	



Fall Detection - Source Code

- Experiments scripts
- Annotated Dataset (1,078 records)
- Pre-trained Word Embeddings

https://github.com/nlp-pucrs/fall-detection



Research Limitations

- Clinical Notes Sampling (natural distribution)
- Self-Attettion Neural Network
- BERT, GPT-2, XLNet Embeddings
- Quality Evaluation (why right, why wrong)



Further Work

Sequence Tagging Task: F-Measure

	LSTM + CRF
Wikipedia	0.55
NILC (broad-domain)	0.67
EHR (specific-domain)	0.80



Further Work

Quality Embeddings - Analogies Task

	Specific	Broad
Wikipedia	1.38 %	79.00 %
NILC (broad-domain)	2.61%	82.38 %
EHR (specific-domain)	2.85 %	0.00 %



Thanks! Questions?

http://www.inf.pucrs.br/ia-saude/

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