



شبكة المعلومات الجامعية
التوثيق الإلكتروني والميكروفيلم

بسم الله الرحمن الرحيم



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جامعة عين شمس

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NAMED ENTITY RECOGNITION FROM BIOMEDICAL TEXT

Thesis submitted as a partial fulfillment of the requirements for the degree of
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Abstract

Named entity recognition is commonly considered an essential task in natural language processing. It represents a major phase in information extraction methodology to discover the named entities referenced in unstructured text and classify them into predefined class labels. Identifying biomedical entities has been recognized as a challenging task in named entity recognition.

In this thesis, the applicability of using structured support vector machine to classify flat and nested biomedical entities combined with the feature selection techniques to enhance the performance of biomedical named entity recognition has been thoroughly investigated. The proposed approach used a combination of various types of features to explore the classification performance in a combination of structured support vector machine as a machine learning technique. These features include linguistic, morphological, orthographical, context, and word representation features. The experimental results showed that the performance of the proposed approach surpassed that produced from other benchmark approaches in extracting the biomedical entities such as genes, proteins, cell lines, cell types, DNAs and RNAs.

Derived by these promising results, we were motivated to explore the effect of different types of features on structured support vector machine performance in extracting the biomedical entities. This was achieved by applying two types of filter-based feature selection techniques. The Chi-squared and ReliefF feature selection techniques were utilized to reduce the feature vector and evaluate the importance of each feature. The reduced

feature vector was then taken as an input to structured support vector machine to extract the biomedical entities. Experimental results revealed that the overall performance of biomedical named entity recognition increased when using the adopted feature selection techniques and removed the features with the least effect on the classes.

Most studies in the literature ignored the nested entities and focused on the extraction of flat named entities only. However, the nested entities are commonly used in real world biomedical applications due to their ability to represent semantic meaning of the named entity. Therefore, the proposed approach was evaluated on the extraction of the nested biomedical entities and showed very promising results compared to those obtained from the benchmark approaches.

Comprehensive experiments were conducted on three popular datasets based on five evaluation metrics namely: recall, precision, F1-measure, Geometric Mean, and Matthews correlation coefficient. Experimental results revealed that the structured support vector machine achieved better performance compared to different approaches in the literature for extraction of both flat and nested entities.

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List of Abbreviations

AI	Artificial Intelligence
ANN	Artificial Neural Network
BA	Boundary Assembly
BCC-NER	Bidirectional Contextual Clues Named Entity Recognition
BERT	Bidirectional Encoder Representations from Transformers
BiLSTM	Bidirectional Long Short-Term Memory
BNER	Biomedical Named Entity Recognition
CNNs	Convolutional Neural Networks
CNN-LSTM	Convolutional Neural Network and Long Short-Term Memory
CRF	Conditional Random Field
DNA	Deoxyribonucleic Acid
FLSTM	Forward Long Short-Term Memory
FN	False Negative
FP	False Positive
G-mean	Geometric Mean
GPRO	Gene and Gene Product
HMMs	Hidden Markov Models
IE	Information Extraction
KeBioLM	Knowledge enhanced Biomedical pretrained Language Model
LSTM	Long Short-Term Memory Networks
MCC	Matthews correlation coefficient
MEMMs	Maximum Entropy Markov Models
ML	Machine Learning
MultiCNN	Multiple Convolutional Neural Networks
NER	Named Entity Recognition
NEs	Named Entities
Nested BNER	Nested Biomedical Named Entity Recognition
NLP	Natural Language Processing

PCA	Principal Component Analysis
POS	Part of Speech
PSO	Particle Swarm Optimization
RNA	Ribonucleic Acid
SPBA	Statistical Principle Based Approach
SSVM	Structured Support Vector Machine
SVMs	Support Vector Machines
TCN	Temporal Convolutional Network
TCN-CRF	Temporal Convolutional Network with a Conditional Random Field
TP	True Positive
UMLS	Unified Medical Language System
WE	Word Embedding
WR	Word Representation

List of Publications

Lobna A. Mady, Yasmine M. Afify, Nagwa L. Badr, “Biomedical Named Entity Recognition Using Structured Support Vector Machine,” International Journal of Computers and Their Applications, Vol. 28, No. 4, pp. 222-229, 2021.

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