

Accepted Manuscript

De-identification of medical records using conditional random fields and long short-term memory networks

Zhipeng Jiang, Chao Zhao, Bin He, Yi Guan, Jingchi Jiang

PII: S1532-0464(17)30222-8
DOI: <https://doi.org/10.1016/j.jbi.2017.10.003>
Reference: YJBIN 2868

To appear in: *Journal of Biomedical Informatics*

Received Date: 31 January 2017
Revised Date: 30 September 2017
Accepted Date: 3 October 2017

Please cite this article as: Jiang, Z., Zhao, C., He, B., Guan, Y., Jiang, J., De-identification of medical records using conditional random fields and long short-term memory networks, *Journal of Biomedical Informatics* (2017), doi: <https://doi.org/10.1016/j.jbi.2017.10.003>

This is a PDF file of an unedited manuscript that has been accepted for publication. As a service to our customers we are providing this early version of the manuscript. The manuscript will undergo copyediting, typesetting, and review of the resulting proof before it is published in its final form. Please note that during the production process errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.



De-identification of medical records using conditional random fields and long short-term memory networks

Zhipeng Jiang*

Chao Zhao*

Bin He

Yi Guan[†]

Jingchi Jiang

School of Computer Science and Technology

Harbin Institute of Technology

Harbin, Heilongjiang, 150001, CHN

HIT.JIANG@HOTMAIL.COM

ZHAOCHAOC@GMAIL.COM

HEBIN_HIT@HOTMAIL.COM

GUANYI@HIT.EDU.CN

JIANGJINGCHI0118@163.COM

Abstract

The CEGS N-GRID 2016 Shared Task 1 in Clinical Natural Language Processing focuses on the de-identification of psychiatric evaluation records. This paper describes two participating systems of our team, based on conditional random fields (CRFs) and long short-term memory networks (LSTMs). A pre-processing module was introduced for sentence detection and tokenization before de-identification. For CRFs, manually extracted rich features were utilized to train the model. For LSTMs, a character-level bi-directional LSTM network was applied to represent tokens and classify tags for each token, following which a decoding layer was stacked to decode the most probable protected health information (PHI) terms. The LSTM-based system attained an i2b2 strict micro- F_1 measure of 0.8986, which was higher than that of the CRF-based system.

Keywords: Protected health information, de-identification, conditional random fields, long short-term memory networks

1. Introduction

The Electronic Health Record (EHR) is the systematized collection of electronically stored health information of patients in digital format [1]. It consists of a large amount of medical knowledge, and is a novel and rich resource for clinical research. A limitation of the large-scale use of EHR is the privacy of information contained in the text. To protect the privacy of patients and medical institutions, the US Congress passed the Health Insurance Portability and Accountability Act (HIPAA) in 1996. HIPAA defines 18 kinds of protected health information (PHI) that must be removed before the EHR can be reused, such as names, all geographic subdivisions smaller than a State, and so on¹. The

*. These authors contributed equally to this work.

[†]. corresponding author

1. <https://www.hipaa.com/hipaa-protected-health-information-what-does-phi-include/> lists all PHIs in the HIPAA

Download English Version:

<https://daneshyari.com/en/article/6927624>

Download Persian Version:

<https://daneshyari.com/article/6927624>

[Daneshyari.com](https://daneshyari.com)