# A de-identification model for clinical notes using deep learning: **Application to Korean language**

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### Background

- The increasing use of electronic health records (EHR) has sparked interest in extracting information from free text within EHR
- Protected health information (PHI) in clinical reports requires de-identification before using the free text
- In South Korea, studies on PHI identification exist, but there is a lack of cross-institutional patient data applications
- Limited prior studies on PHI de-identification in medical big data have been conducted in Korea, resulting in inadequate development in this area
- Our objective was to identify the PHI list and fine-tune the BERT model for developing a PHI deidentification tool

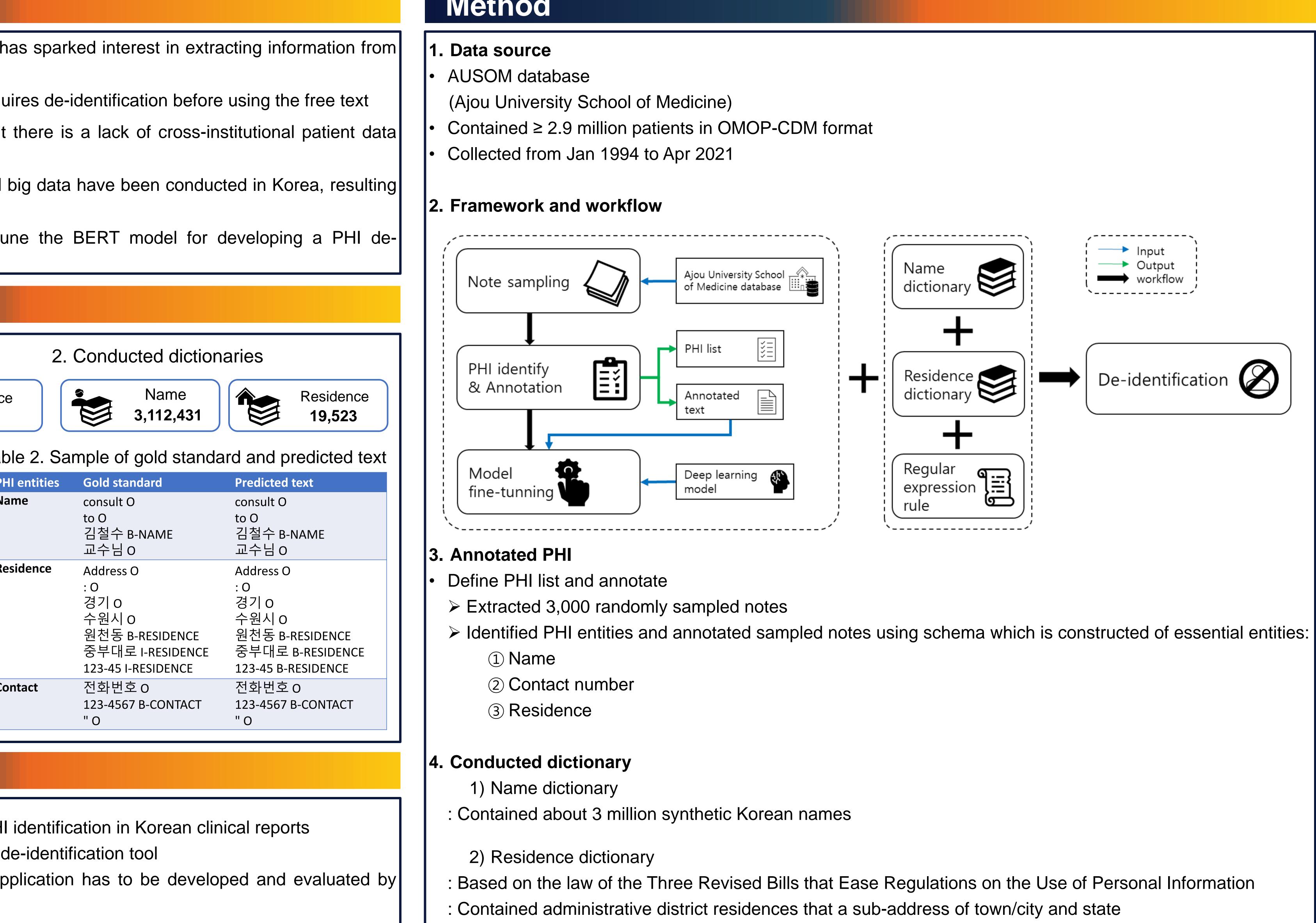
### Results

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### Conclusions

- In this study, we successfully fine-tuned a model for PHI identification in Korean clinical reports
- This model can serve as a foundation for developing a de-identification tool
- based on this model, in further study, the software application has to be developed and evaluated by external validation in various hospital database

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### Method

