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History Question Classification and Representation for Chinese Gaokao

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We proposed a question representation based on entity labeling and question classification for a automatic question answering system of Chinese Gaokao history question. A CRF model is used for the entity labeling and SVM/CNN/LSTM models are tested for question classification.

The contributions of our work: **D**atasets Annotations

Figure 1. Procedures of answering history material questions

We use a corpus which built with questions in past Chinese Gaokao examination papers of history subject and exercises. And manually annotated their question types and question entities.

Question Classification

For the purpose of comparison, we first use a rulebased method which use regular expressions. We choose SVM as our baseline. Then CNN and LSTM are used which works better than SVM.

Question Entity Recognition

We apply CRF which provide a good performance.



What we are going to do in this paper is classifying the questions of Gaokao history subject, which is the first step of our question answering system as shown in figure 1.

Table II Classes of the question

Datasets Annotations			Table II Classes of the question		
			Туре	Description	Example
Table I Label of question entity					据材料三,简析秦国民风变化 带来的影响。
Туре	Description	Example		Compare events or people. Common, difference or both	根据上述材料,比较李鸿章和 郭嵩焘主张的异同。
Material Event	The index of material The history event	材料一, 材料 辛亥革命,五四运动		Comment on the people or the events	结合材料二和所学知识,评价 严复的思想主张。
Focus Person	Key point of question Person names	意义,背景 伏尔泰,孔子	Open-ended		依据上述材料,你认为当今世
Location	Locations	中国,欧洲	Judgement	Judge if it is right or which one is most likely to be the answer	材料二中说"北军的胜利,美国的工业大局已定"是否正确?
Time Organizatior		唐朝,明,19世纪 巴黎公社,联合国		Describe the cause of event or the effect of the event	据材料三,简析秦国民风变化 带来的影响。

Table I shows the 7 kinds of question entities. These are the most import parts with which we can search in the knowledge base more accurately.

Table II shows the 6 classes of the history questions. The reason why we classify in this way is that most of the questions need some skills to answer which can help to get a high score in the exam.

Table III Accuracy of question classification on history questions datasets

Method	Accuracy	Macro-F1
Rule-based	81.33%	63.35%
SVM	85.56%	81.16%
CNN	88.84%	83.31
LSTM	90.25%	87.43%

Table III shows that neural networks outperform the SVM method we implemented on the corpus. All of the above machine learning methods work better than rule-based method. meanings, it can be Qin dynasty or country.

Table IV Accuracy of question entity recognition

Label	Precision	Recall	F1
Material	97.16%	87.05%	87.43%
Event	80.49%	84.62%	82.50%
Focus	83.78%	86.11%	84.93%
Person	89.90%	86.41%	88.12%
Location	86.55%	86.55%	86.55%
Time	82.93%	80.00%	81.44%
Organization	86.49%	84.62%	82.50%

The main difficulty is that most of the history entities are new to the tokenizer. Meanwhile, some words such as ``秦'' has multiple

