



## Database Group

## Scalable Graph Analysis over Intel Xeon Phi Coprocessors

The structural graph clustering method SCAN is successfully used in many applications since it detects not only densely connected nodes as clusters but also extracts sparsely connected nodes as hubs or outliers (Fig. 1). However, it is difficult to apply SCAN to large-scale graphs since SCAN needs to evaluate the density for all adjacent nodes included in the graph. In this work, so as to address the above problem, we present a novel algorithm SCAN-XP that performs on Intel Xeon Phi coprocessors. We designed SCAN-XP to make best use of many cores in the Intel Xeon Phi by employing the following approaches: First, SCAN-XP avoids the bottlenecks that arise from parallel graph computations by providing good load balances among the cores. Second, SCAN-XP effectively exploits 512 bit SIMD instructions implemented in each core to speed up the density evaluations. As a result, SCAN-XP runs approximately 100 times faster than SCAN; for the graphs with 100 million edges, SCAN-XP is able to perform in a few seconds (Fig. 2). **Table. 1: Real-world Dataset** 

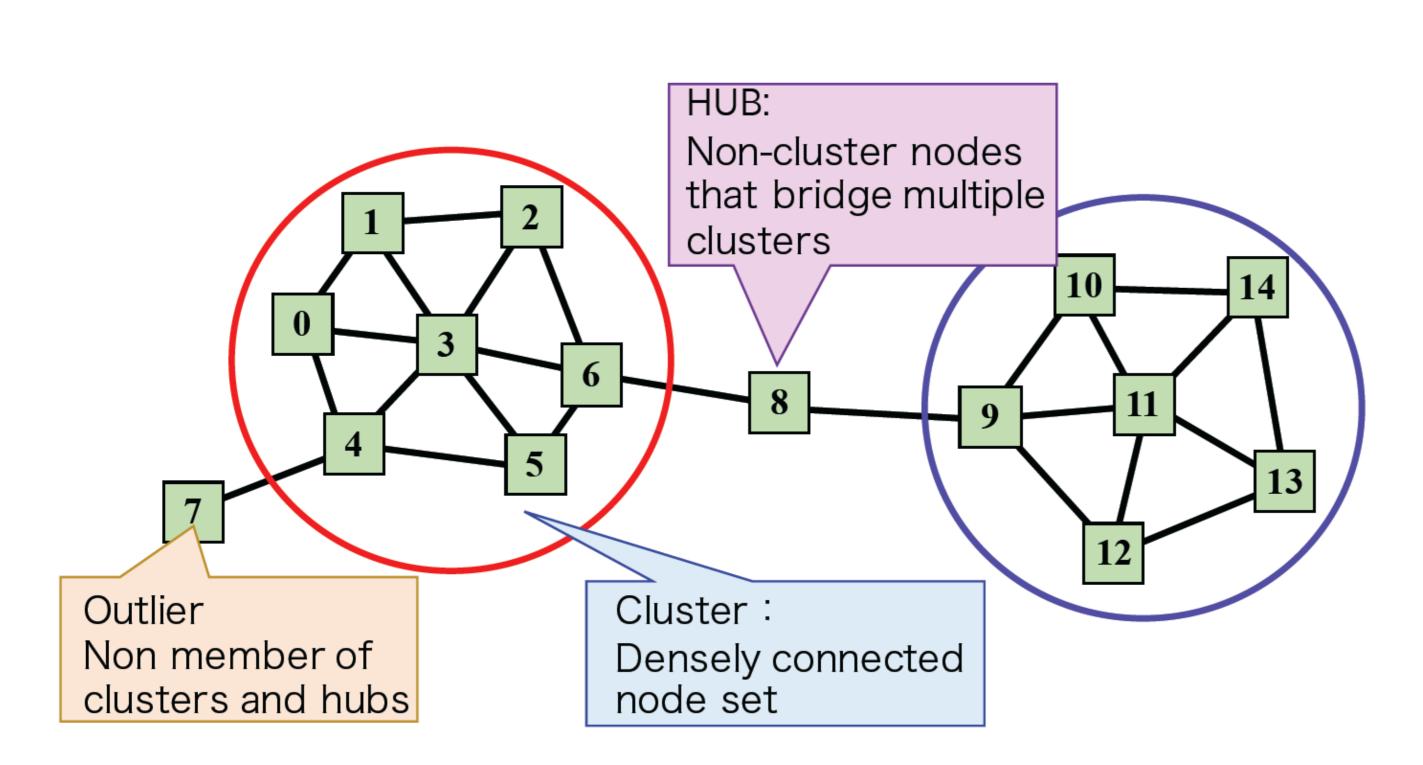
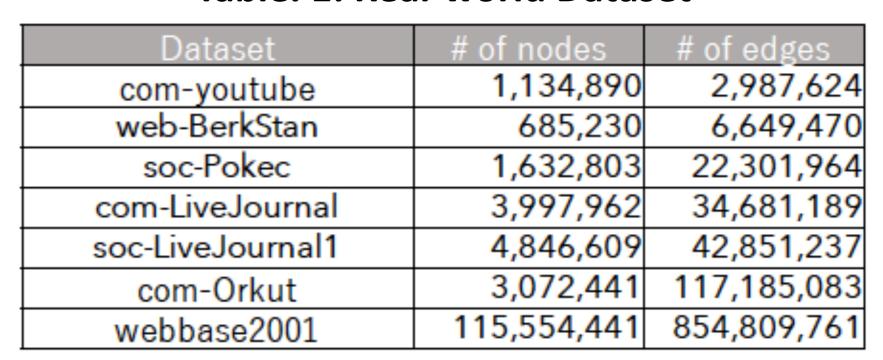


Fig. 1: Structural Graph Clustering SCAN



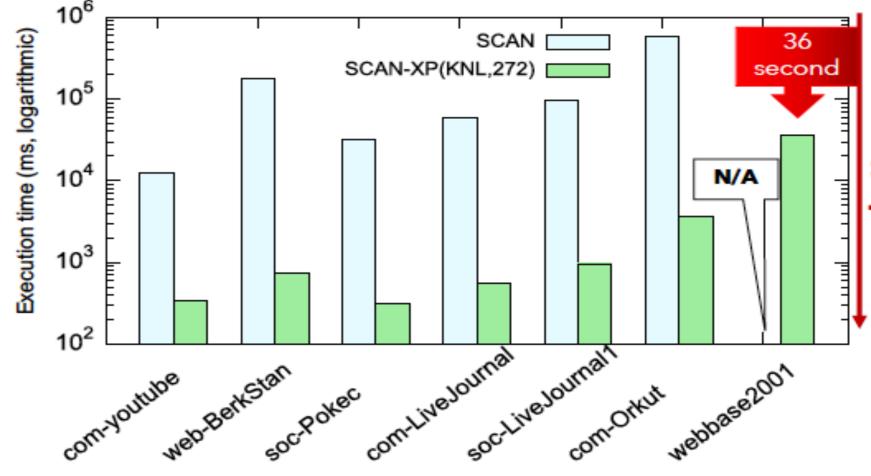


Fig. 2: Overall performance

## Noise-robust sleep stage scoring for mice using deep learning & big data

Sleep stage scoring for mice is one of the most basic analyses in sleep research. However, this analysis takes so much time and effort, which causes degradation of research efficiency. Although automated scoring methods have been proposed, they do not have enough robustness against noise and are not suited for research purpose. To develop noise-robust scoring method, we take the following approaches.

- Employing convolutional neural network (CNN) & long short-term memory (LSTM), which enable us to capture the features of not only biological signals but also their noise.
- Training the model using noisy biological signals obtained from over 3000 mice.

The proposed method can score sleep stages with accuracy of more than 95%, even if the noisy signals are inputted. This is enough practical to use in sleep research.

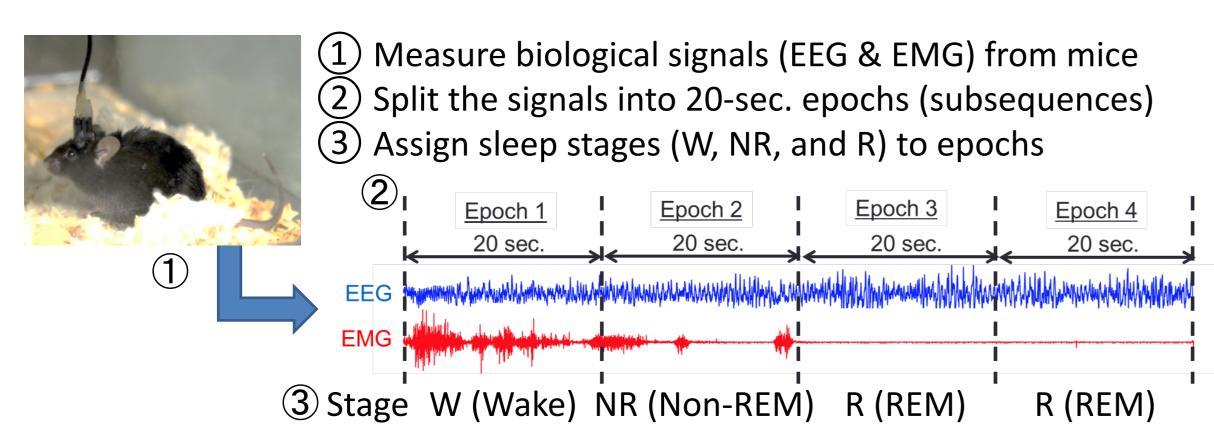


Fig. 3: Procedure of sleep stage scoring

Table 2: Feature of each stage

Table 2. realare of each stage		
Stage	Peak Freq. of EEG	Amplitude of EMG
W	7-11 Hz	Large
NR	1-6 Hz	Small
R	7-11 Hz	Smallest

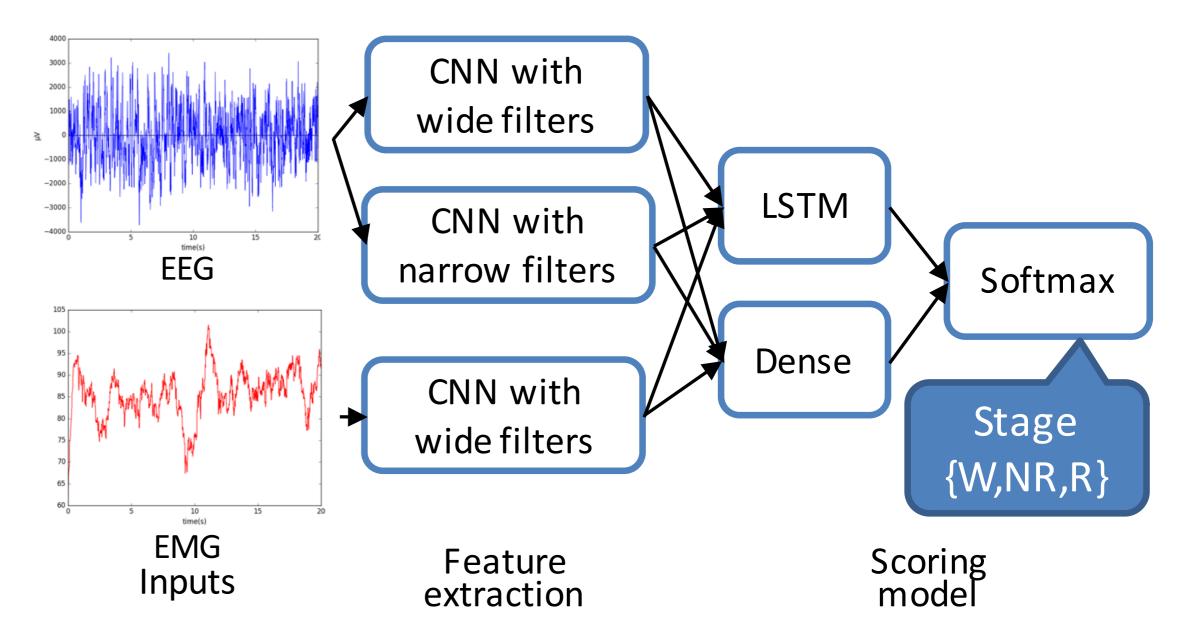


Fig. 4: Structure of the proposed system