A Spatial-Temporal Graph Convolutional Networks-based **3**rd Approach for the OpenPack Challenge 2022

Shurong Chai, Jiaqing Liu, Rahul Jain, Yinhao Li, Tomoko Tateyama, Yen-Wei Chen

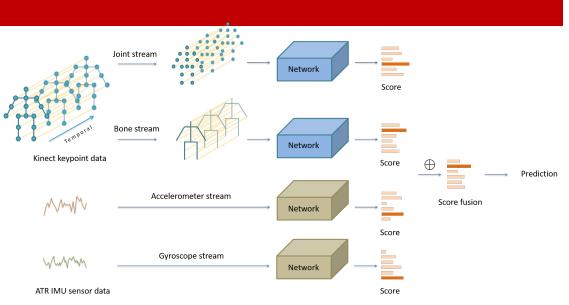


RITSUMEIKAN

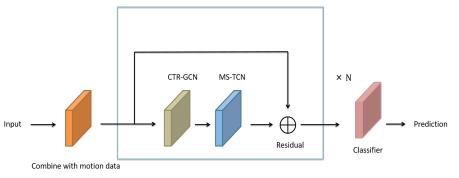
Method Overview

Novelty:

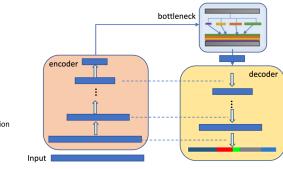
- Multimodal Feature extraction => joint, bone, accelerometer, gyroscope
- Motion data changes dramatically from one action to another=> Combine normal and motion data
- Capture the long-range dependencies among temporal dimensions =>propose a multi-scale temporal convolution network employing largesize kernels
- Over segmentation problem => smoothing loss



Keypoint data stream



Sensor data stream



Accelerometer, Gyroscope stream

Capture information at different resolutions

Add a classifier to predict the action boundary[1]

Conv 1 \times 1

Conv 17 \times

 $[T, N, \frac{C_{in}}{2}]$

 $[T, N, \frac{C_{in}}{8}]$

Temporal feature extraction

 $x[T, N, C_{in}]$

Conv 1 × 1

Conv 17 >

 $[T, N, \frac{C_{in}}{8}]$ $[T, N, \frac{C_{in}}{8}]$

Concatnate

dilation

 $[T, N, \frac{C_{in}}{2}]$

 $Conv 1 \times 1$

Conv 17 × 1

dilation

 $[T, N, \frac{C_{in}}{2}]$

 $-x_{t-2}$

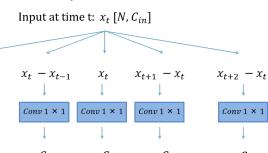
 $[N, \frac{C_{in}}{9}]$

Ground-truth:

Prediction :

Motion-aware input

Small-scale temporal difference Larger-scale temporal difference



 $x_t [N, C_{in}]$

 $L_{keypoint} = L_{CrossEntropy} + L_{TMSE}$

Loss functions

Over segmentation problem [3]

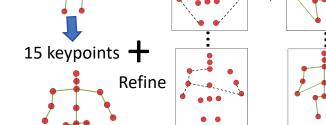
 $[N, \frac{C_{in}}{8}]$ $[N, \frac{C_{in}}{2}]$ $[N, \frac{C_{in}}{8}]$ $[N, \frac{C_{in}}{2}]$ N: Number of joint Concatnate

C: Input channels T:Video length

17 keypoints

T: Video length

Spatial feature extraction



Channel-wise topology refinement graph convolution[2]

$x[T, N, C_{in}]$ Multi-scale temporal convolution

F1 score = 0.924

References

Results

C: Number of classes [1] Singhania, D., Rahaman, R., & Yao, A. (2021). Coarse to fine multi-resolution temporal convolutional network. arXiv preprint arXiv:2105.10859.

[2] Chen, Yuxin, et al. "Channel-wise topology refinement graph convolution for skeleton-based action recognition." Proceedings of the IEEE/CVF International Conference on Computer Vision. 2021. [3]Farha, Y. A., & Gall, J. (2019). Ms-tcn: Multi-stage temporal convolutional network for action segmentation. In Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition (pp. 3575-3584).

 $L_{\text{TMSE}} = \frac{1}{TC} \sum_{t,c} \tilde{\Delta}_{t,c}^2 \qquad \tilde{\Delta}_{t,c} = \begin{cases} \Delta_{t,c} \colon & \Delta_{t,c} \leq \tau \ y_{t,c} \colon \text{Probability of} \\ \tau \colon \text{otherwise class c at time t} \end{cases}$ $\tau = 16$ $\Delta_{t,c} = |\log y_{t,c} - \log y_{t-1,c}|$