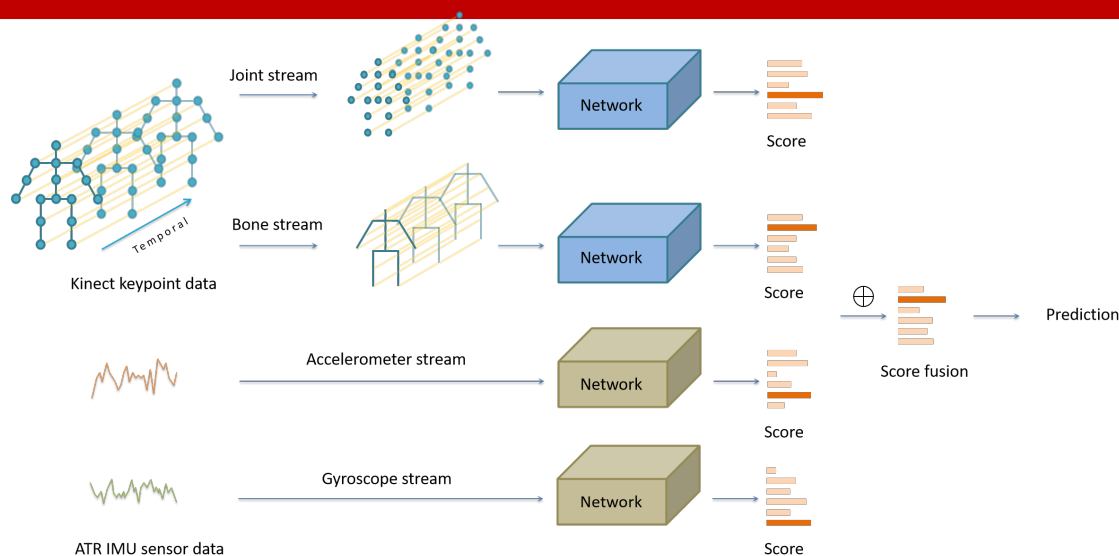


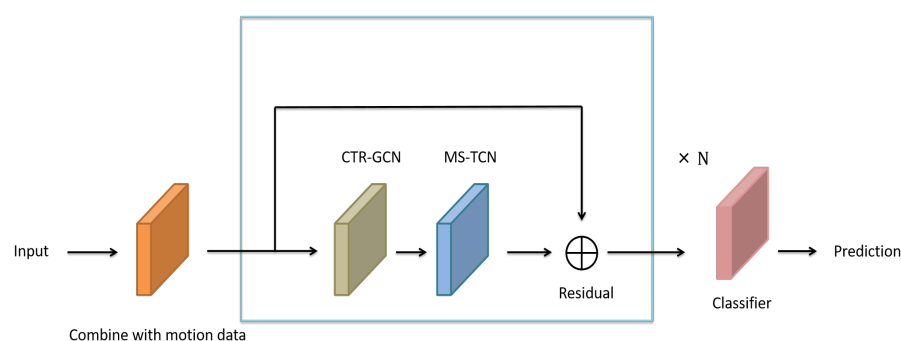
Method Overview

Novelty:

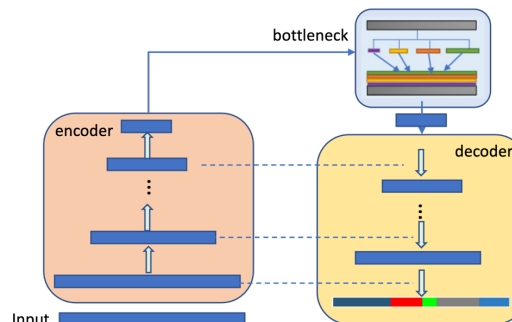
- Multimodal Feature extraction
=> joint, bone, accelerometer, gyroscope
- Motion data changes dramatically from one action to another=> Combine normal and motion data
- Capture the long-range dependencies among temporal dimensions => propose a multi-scale temporal convolution network employing large-size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



Keypoint data stream



Sensor data stream



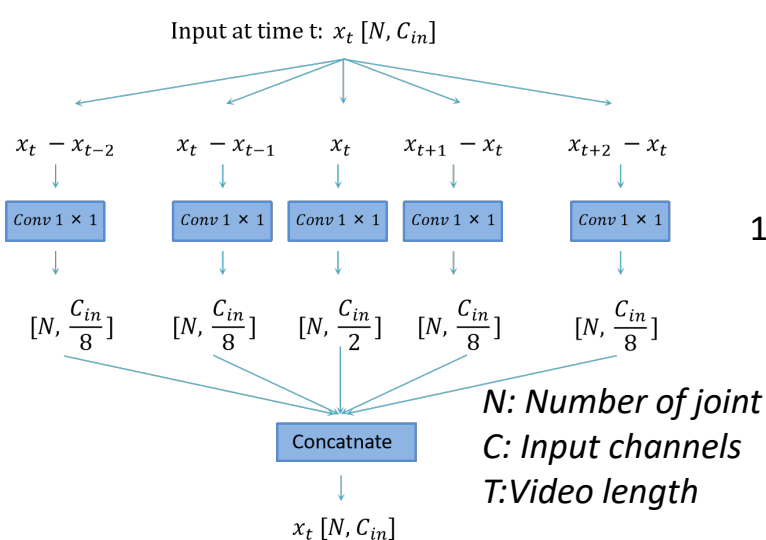
Accelerometer, Gyroscope stream

Capture information at different resolutions

Add a classifier to predict the action boundary[1]

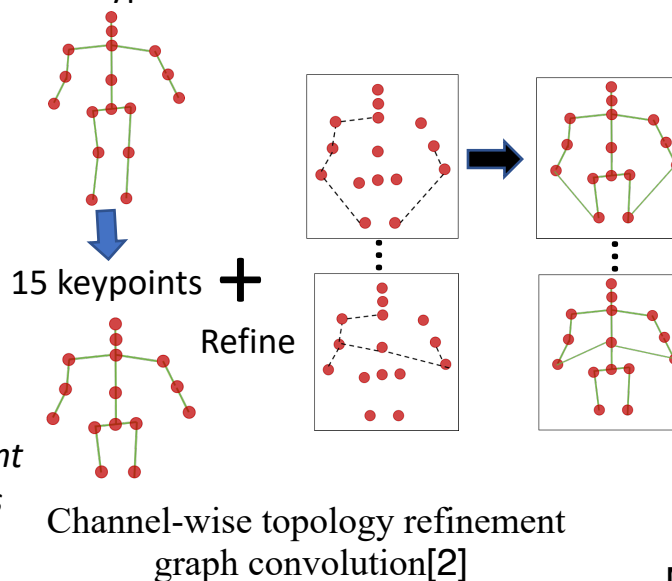
Motion-aware input

Small-scale temporal difference
Larger-scale temporal difference

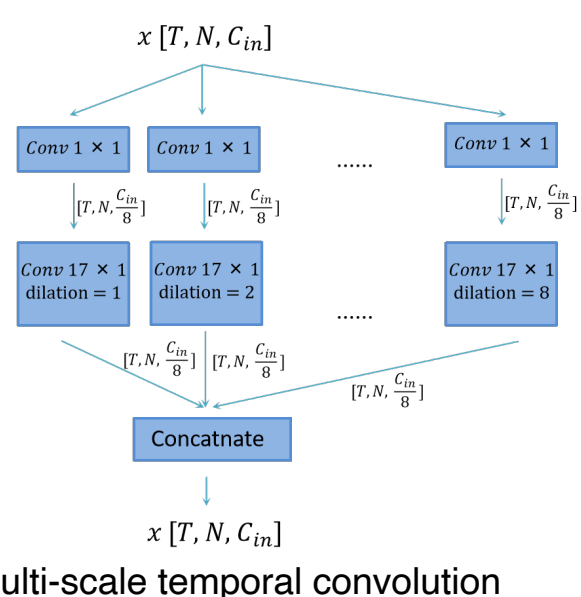


Spatial feature extraction

17 keypoints



Temporal feature extraction



Loss functions

Over segmentation problem [3]

Ground-truth:

Prediction:

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

$$L_{TMSE} = \frac{1}{TC} \sum_{t,c} \tilde{\Delta}_{t,c}^2 \quad \tilde{\Delta}_{t,c} = \begin{cases} \Delta_{t,c} & \Delta_{t,c} \leq \tau \\ \tau & \text{otherwise} \end{cases}$$

$$\Delta_{t,c} = |\log y_{t,c} - \log y_{t-1,c}|$$

T : Video length

C : Number of classes

$y_{t,c}$: Probability of class c at time t

$\tau = 16$

Results

$F1 \text{ score} = 0.924$

References

- [1] Singhanian, 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).