

# Recognizing 3D actions with shallow convolutional neural networks and quad-joint relative volume maps

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## Abstract

In this study, we present a new set of feature maps for use with a Circular convolutional neural network (CCNN) that overcomes the shortcomings of the prior maps and allows for superior pattern discrimination. By leveraging the local relationships among joint movements represented by three-dimensional quadrilaterals constructed for every conceivable set of four joints, these novel characteristics calculate the volumes of these time-varying quadrilaterals. As a result, they produce color-coded images known as spatio temporal quad-joint relative volume maps (QjRVMs).

## 1.Introduction

Compared to other works already in existence, the average recognition accuracy of the prior one is 90.32%. By expanding the number of joints included in the research from 2 or 3 to 4, and then running the results through a circular convolution neural network (CCNN), this work provides a deeper dive into the issues that have arisen in prior research. The design, findings, and potential evaluations are all laid out in this study. The figure.1 depicts the proposed comparison and joint selection method. Multi-tensor images, which are the physical embodiment of features extracted manually from 3D spatiotemporal activity data, must be used as input to CNNs. Joint, edge, angle, surface, and distance maps (JDMs) and their variants are only a few examples of existing feature maps that can be used singly or in tandem. Therefore, color-coded maps derived from these characteristics are insufficient for capitalizing on multi-joint linkages in critical action sequences. When compared to relative volume feature maps, feature maps tend to be weaker. The proposed features are computed volumes on a set of four joints that create a time-varying 3D quadrilateral over the action sequences in a video series. During actions, quadrilaterals are constructed from all conceivable sub-joint orientations. The collected quadrilateral volumes from a 3D video stream are used to generate color-coded feature maps known as quad joint relative volume

(QjRVM). In Figure 1(j), we contrast the currently used features.1. (a)-(i) with our proposed volume features.

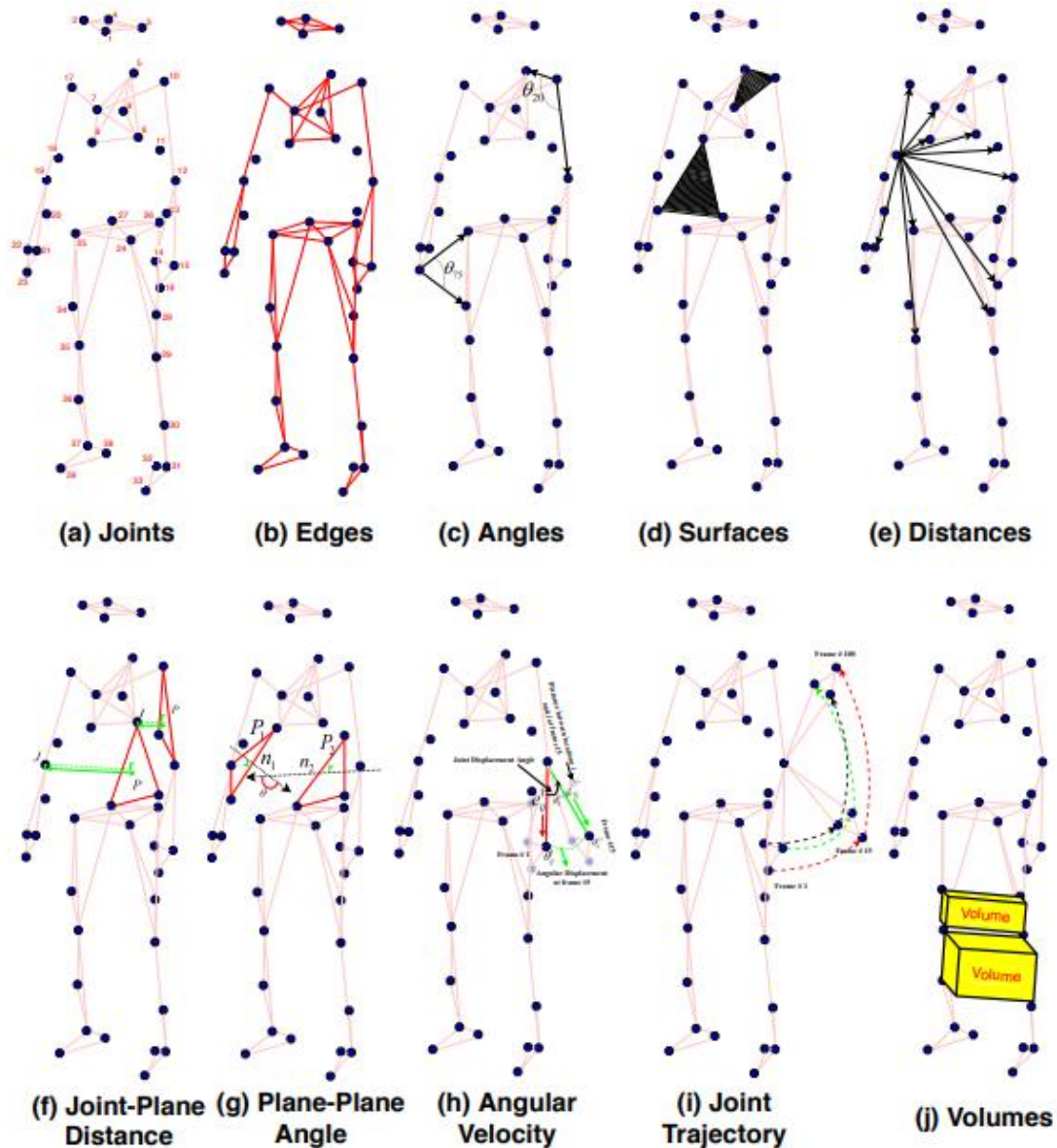


Fig. 1: Feature models used for 3D action recognition. (a) Joints, (b) Edges, (c) Angles, (d) Surfaces, (e) Distances and (f) Joint – Plane distance, (g) Plane – Plane angle, (h) Angular velocity, (i) Joint trajectory and (j) Relative volume features.

To create a 3D tensor image, a proportionate coloring scheme is used to the retrieved features. Visualize a temporal and spatial progression of events with the help of color-coded image tensors. Color-coded maps from different models are compared in Figure 2.

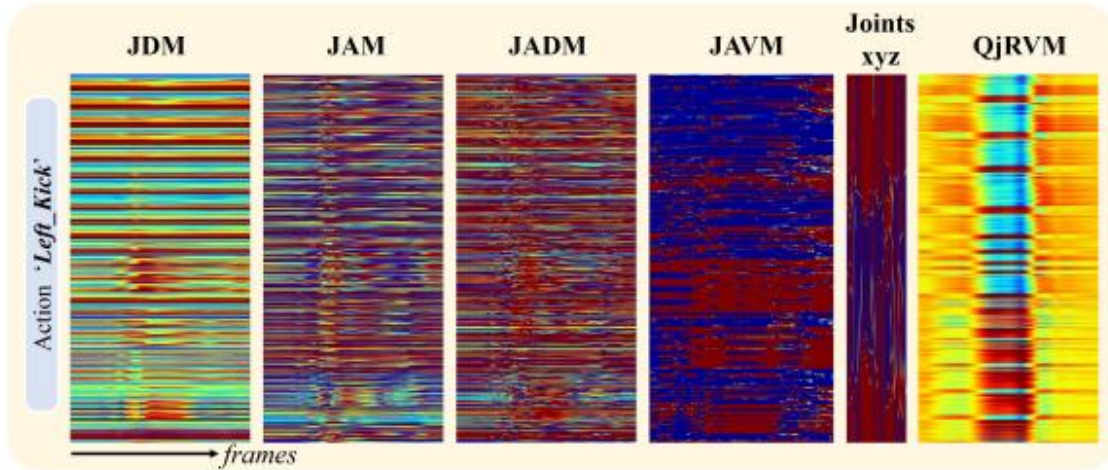


Fig. 2: Showing a formal comparison among color-coded feature maps of our proposed relative quad joint relative volume and the earlier joint feature maps.

## 2. Model for Identifying Human Actions, Version

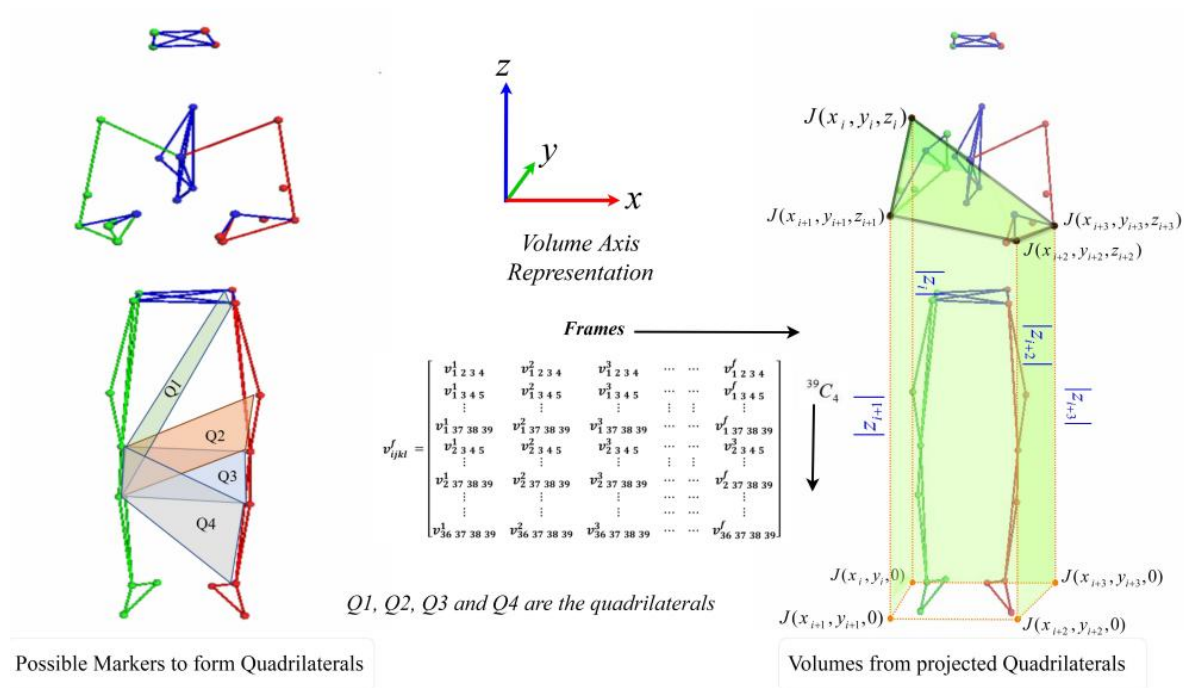


Fig. 3: Construction of relative QJRVMs are designed

As the skeleton rotates and flexes in a 3D film, the resulting shapes and orientations of the joints reflect this relationship. Using these versions, we can build the QjRV matrices to represent the relative volumes of four joints. To find the volume of a quadrilateral given its coordinates  $\{(x_i, y_i, z_i), (x_{i+1}, y_{i+1}, z_{i+1}), (x_{i+2}, y_{i+2}, z_{i+2}), (x_{i+3}, y_{i+3}, z_{i+3})\}$  of its vertices, this procedure is as follows: The relative QJRVMs are designed in figure.3

### 3. Conclusion

In this paper, we offer a new method for visualizing 3D motion capture data. In addition, 102 lessons covering five topics were incorporated into the project's preexisting data set. The technique involves the development of computationally effective QjRVMs for all skeleton joints, which are then used to generate color-coded images indicating spatiotemporal variations in 3D sequences of motion. These QjRVMs disclose relative local changes in joint positions during activity as a color shift in the pixels, whereas previous maps are unrelated and more focused on the location of joints with distance information. In order to extract patterns from these QjRVMs, a DNN architecture called CCNN was developed for this study. During training, CCNN's four streams, each having its own set of filters that cycle cyclically from bottom to top, are used to detect different spectral properties.

### References

1. C. Li, Y. Hou, P. Wang, and W. Li, "Joint distance maps based action recognition with convolutional neural networks," *IEEE Signal Processing Letters*, vol. 24, no. 5, pp. 624–628, 2017.
2. M. Z. Uddin, N. D. Thang, and T.-S. Kim, "Human activity recognition via 3-d joint angle features and hidden markov models," in *Image Processing (ICIP), 2010 17th IEEE International Conference on*. IEEE, 2010, pp. 713–716.
3. J. Liu, G. Wang, P. Hu, L.-Y. Duan, and A. C. Kot, "Global context-aware attention lstm networks for 3d action recognition," in *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, 2017, pp. 1647–1656.
4. E. K. Kumar, P. V. V. Kishore, M. T. K. Kumar, D. A. Kumar, and A. S. C. S. Sastry, "Three-dimensional sign language recognition with angular velocity maps and convolved feature ResNet," *IEEE Signal Processing Letters*, vol. 25, no. 12, pp. 1860–1864, dec 2018.
5. D. Wu and L. Shao, "Leveraging hierarchical parametric networks for skeletal joints based action segmentation and recognition," in *Proceedings of the IEEE conference on computer vision and pattern recognition*, 2014, pp. 724–731
6. S. Abbas, M. Benchohra, J.E.Lagreg and Y.Zhou, A survey on Hadamard and Hilfer fractional differential equations: Analysis and stability, *Chaos, Solitons and Fractals*, 102, 47 – 71, 2017.

7. Halim, N.A. Haq, R.U. Noor, N.F.M. (2017), Active and passive controls of nano particles in Maxwell stagnation point flow over a slipped stretched surface, *Meccanica*, 52(7), 1527–1539.
8. Vijaya, N. Hari Krishna, Y. Kalyani, K. and Reddy, G. V. R. (2018), Soret and radiation effects on an unsteady flow of a Casson fluid through porous vertical channel with expansion and contraction, *Frontiers in Heat and Mass Transfer*, 11, doi:10.5098/hmt.11.19
9. Vijaya, N. Madhavi, M. R. and Krishna, Y. H. (2018), Boundary layer flow of a mixed convective nanofluid over a vertical circular cylinder under the influence of magnetic field, heat radiation and external surface temperature, *International Journal of Mechanical and Production Engineering Research and Development*, 8, 411-420
10. Reddy, G. V. R. and Krishna, Y. H. (2018), Numerical solutions of unsteady MHD flow heat transfer over a stretching surface with suction or injection, *Fluid Dynamics and Materials Processing*, 14(3), 213-222.