

Article

GeoSparseNet: A Multi-Source Geometry-Aware CNN for Urban Scene Analysis

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Abstract: The convolutional neural networks (CNNs) functioning on geometric learning for the urban large-scale 3D meshes are indispensable because of their substantial, complex, and deformed shape constitutions. To address this issue, we proposed a novel Geometry-Aware Multi-Source Sparse-Attention CNN (GeoSparseNet) for the urban large-scale triangular mesh classification task. GeoSparseNet leverages the non-uniformity of 3D meshes to depict both broad flat areas and finely detailed features by adopting the multi-scale convolutional kernels. By operating on the mesh edges to prepare for subsequent convolutions, our method exploits the inherent geodesic connections by utilizing the Large Kernel Attention (LKA) based Pooling and Unpooling layers to maintain the shape topology for accurate classification predictions. Learning which edges in a mesh face to collapse, GeoSparseNet establishes a task-oriented process where the network highlights and enhances crucial features while eliminating unnecessary ones. Compared to previous methods, our innovative approach outperforms them significantly by directly processing extensive 3D mesh data, resulting in more discerning feature maps. We achieved an accuracy rate of 87.5% when testing on an urban large-scale model dataset of the Australian city of Adelaide.

Keywords: deep learning; 3D meshes; urban-scale; remote sensing; Geometry-Aware; attention



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1. Introduction

The 3D scene or shape analysis of a concoction of dense, complex, curved surfaces, or irregular geometries of meshes in large urban environments including buildings, trees, cars, and other elements has been made possible by recent developments in 3D computer vision and photogrammetry [1]. To enable the use of these meshes in a variety of applications, such as smart urban planning, navigation systems, virtual reality, radiation estimation, noise modeling, and photovoltaic perspective, it is essential to extract semantic information from the mesh models [2,3].

For semantic classification of 3D mesh data, a trivial number of machine learning-based algorithms are available; and their primary focus is to process 3D point clouds [4,5].