

# FoGMesh: 3D Human Mesh Recovery in Videos with Focal Transformer and GRU

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#### Introduction

#### **➤** Motivation:

- High-precision 3D human body recovery in the wild is a challenging task, especially in high-speed motion scenes.
- The attention weight of small-scale body parts is small in the reconstruction process, resulting in large deviation of the recovery results.

#### Contributions:

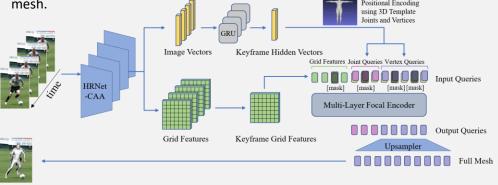
- A new method, FoGMesh, taking the advantage of the focal attention mechanism to model both local and global interactions for 3D human mesh recovery.
- We propose the CAA module that merges multiscale features and effectively amplifies the attention weights of small-scale body parts.

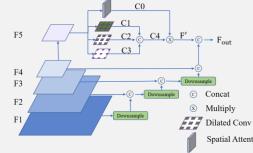
## **Our Results on 3DPW**

| Method     | 3DPW  |        |           |
|------------|-------|--------|-----------|
|            | MPVE↓ | MPJPE↓ | PA-MPJPE↓ |
| HMR        | -     | -      | 81.3      |
| SPIN       | 116.4 | -      | 59.2      |
| RSC-Net    | -     | 96.4   | 59.0      |
| Pose2Mesh  | -     | 89.2   | 58.9      |
| VIBE       | 99.1  | 82.0   | 51.9      |
| METRO      | 88.2  | 77.1   | 47.9      |
| Graphormer | 87.7  | 74.7   | 45.6      |
| Ours       | 85.2  | 74.1   | 45.5      |

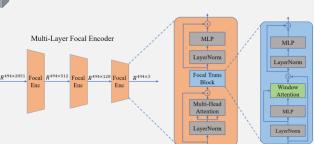
## **The Proposed Method**

Our framework takes a temporal sequence as input to obtain grid features and global feature vectors using a CNN. The global vectors are fed into the GRU encoder to obtain keyframe hidden vectors. The keyframe grid features and keyframe global vectors are tokenized and fed into the MFE for 3D human mesh.





The MFE consists of three focal encoder blocks with the same number of input tokens. We combine finegrained local and coarsegrained global interaction with the proposed MFE encoder module.



concatenating.

The feature map F5 is processed by

the spatial attention module and the dilated convolution with different

rates. Multi-scale feature maps are

fused through downsampling and

#### **Reconstruction Results**



Input METRO Graphormer Attn. Vis. FoGMest

↑ Comparison of prediction results on high-speed

motion blurred images

↑ Prediction results on the clear original images

# **Acknowledgements**

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## References

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- [2] Kevin Lin, Lijuan Wang, and Zicheng Liu. Mesh graphormer. In Proceedings of the IEEE/CVF International Conference on Computer Vision, pages 12939–12948, 2021.