



## Action Recognition using 2D/3D Skeleton Sequences

**Level:** Bachelor/Master (1-2 students possible)

**Duration:** 3 months

**Start:** By agreement

**Mentor:** Milos Petrovic

**Institution:** ETF Robotics

**Overview and Technology:** This project addresses **action recognition** using **2D and 3D pose sequences** extracted from video. Instead of recognizing actions directly from pixels, the system uses a pose estimator to obtain a skeleton time series, reconstructs 3D joints, and then applies spatio-temporal deep models (GCNs, 3D CNNs, or transformers) to classify actions. Students will build an end-to-end pipeline (pose extraction -> skeleton normalization -> training/inference), evaluate on public datasets, and compare skeleton-based methods against RGB+D baselines in terms of accuracy, speed, and robustness to appearance changes.

Platforms / hardware	Software & tools
<ul style="list-style-type: none"><li>• PC Workstation (GPU recommended)</li><li>• Public video datasets</li><li>• RGB+D camera</li></ul>	<ul style="list-style-type: none"><li>• Python (NumPy, OpenCV)</li><li>• PyTorch + OpenMMLab (MMPose, MMAction2 / PySKL)</li><li>• PyTorchVideo (optional RGB baseline)</li></ul>
Project options (projects can be modified based on student interests)	
<ul style="list-style-type: none"><li>• Extract 2D skeletons with MMPose and train MMAction2/PySKL skeleton models (ST-GCN, PoseC3D, etc.)</li><li>• Lift to 3D skeletons (VideoPose3D/MeTRAbs) and benchmark 2D vs 3D recognition</li><li>• Speed-up and deploy: real-time webcam inference, model compression/quantization, lightweight backbones</li></ul>	
Expected outcomes	Recommended background
<ul style="list-style-type: none"><li>• Literature review</li><li>• Project code and documentation/video</li><li>• Final report in IEEE research paper form</li></ul>	<ul style="list-style-type: none"><li>• Digital image processing basics and camera geometry (2D/3D coordinates, normalization)</li><li>• Python programming (NumPy/OpenCV) and basic software engineering</li><li>• Basics of machine learning / deep learning (PyTorch)</li></ul>
Literature	
<ul style="list-style-type: none"><li>• MMAction2 model zoo: <a href="https://mmaction2.readthedocs.io/en/latest/model_zoo/recognition.html">mmaction2.readthedocs.io/en/latest/model_zoo/recognition.html</a> and <a href="https://mmaction2.readthedocs.io/en/latest/model_zoo/skeleton.html">mmaction2.readthedocs.io/en/latest/model_zoo/skeleton.html</a></li><li>• MMPose for keypoint extraction; alternatives: MediaPipe Pose Landmarker and OpenPose</li><li>• ST-GCN paper; PySKL toolbox; PyTorchVideo (optional RGB baselines)</li><li>• OpenPose / MediaPipe for pose extraction</li></ul>	