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Bangladesh University of Business and Technology
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CSE 498A: Literature Review Records

Student's Id and Name	22234103382, Umma Sumaiya Riya
Capstone Project Title	High-Efficiency Micro-Expression Recognition for Automated Human Behavior Analysis Systems.
Supervisor Name & Designation	Md. Mijanur Rahman, Assistant Professor
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Aspects	<p>Paper # 25 Geometrical Facial Expression Recognition Approach Based on Fusion CNN-SVM [Published Year: 2024, Publisher: International Journal of Intelligent Engineering and Systems]</p>
Problem Statement	Facial Expression Recognition (FER) remains challenging due to noise, illumination changes, pose variations, and the inability of single-feature methods to accurately capture subtle emotional changes. Appearance-based models struggle with texture variations, while geometric-only methods fail to generalize across subjects. The authors aim to develop a FER system that robustly combines shape (geometric) and texture (appearance) information to improve accuracy, especially with small datasets like CK+ and JAF
Key Contributions	(i) Introduced a hybrid Fusion CNN-SVM architecture combining appearance features from 2D-CNN and geometric features from β -skeleton-based landmark modeling. (ii) Proposed a geometry extraction method using 68 facial landmarks grouped into seven facial regions and transformed into discriminative geometric vectors.

	<p>(iii) Utilized a 1D-CNN + 2D-CNN dual-stream feature learning mechanism.</p> <p>(iv) Replaced Softmax with L2-SVM for improved classification on small datasets.</p> <p>(v) Achieved competitive results on CK+ and JAFFE benchmarks.</p>
Methodology/Theory/Framework	<ul style="list-style-type: none"> ○ The authors built a two-branch CNN model. ○ Geometric branch: detects 68 facial landmarks, generates β-skeleton geometric features, and processes them with a 1D-CNN. ○ Appearance branch: preprocesses face images (crop, resize, equalize) and extracts texture features using a 2D-CNN. ○ Fusion & classification: both feature streams are merged, and an L2-SVM classifier is used instead of Softmax for better performance on small datasets.
Software/Tools/Setup Details	<ul style="list-style-type: none"> <input type="checkbox"/> Programming Language: Python 3.6 <input type="checkbox"/> Libraries: TensorFlow/Keras, Dlib for landmarks <input type="checkbox"/> Hardware: Intel Core i5 CPU, 4GB RAM <input type="checkbox"/> Image size: 50×50 pixels <input type="checkbox"/> Training parameters: Adam optimizer, LR = 0.0001, batch size = 32, 50 epochs <input type="checkbox"/> Data Augmentation: Rotation ($\pm 15^\circ$), flipping, blur, sharpening, mirroring
Test/Experiment Analysis	<p>The authors performed experiments on CK+ and JAFFE datasets with an 80/20 train-test split.</p> <p>They evaluated models on accuracy, precision, recall, and F1-score. Comparisons were made between:</p> <ul style="list-style-type: none"> ● baseline CNN ● geometric-only features ● appearance-only features ● proposed Fusion CNN-SVM model <p>The fusion model outperformed all other approaches, showing improved recognition of happy and surprise emotions, while challenging emotions like fear and disgust showed moderate accuracy. Accuracy:</p> <ul style="list-style-type: none"> ● CK+ dataset: 96.19% ● JAFFE dataset: 89.23%

Test Data/Dataset Source	<p>CK+ Dataset: 593 facial expression sequences, 7 emotion classes, frontal images; a standard FER benchmark.</p> <p>JAFFE Dataset: 213 images of 10 subjects, 7 expressions; collected in a controlled lab environment.</p>
Final Result (Assessment Criteria Wise)	<p>The proposed method demonstrated strong performance in combining appearance and geometric features.</p> <p>The model achieved high accuracy, better generalization on small datasets, improved robustness against pose/illumination variations, and superior results compared to traditional CNN and handcrafted feature methods. The SVM classifier contributed to stable performance and reduced overfitting..</p>
Limitations (List the limitations the authors mentioned in the article)	<p>1)Performance decreases for highly similar expressions such as fear and disgust.</p> <p>2)Dependent on accurate landmark detection; β-skeleton fails if landmarks are wrong.</p> <p>3)Tested on only two small datasets; not evaluated in real-world, unconstrained environments.</p> <p>4)No temporal information (video sequences) included.</p>
Final Summary	<p>The paper presents a hybrid Fusion CNN-SVM framework that integrates geometric features from β-skeleton landmark modeling with appearance features from 2D-CNN. A dual-branch CNN architecture is used to extract both shape and texture-related patterns, which are later fused and classified using an L2-SVM classifier. Experiments conducted on CK+ and JAFFE datasets show that the proposed method achieves accuracy values of 96.19% and 89.23%, outperforming baseline CNN and traditional handcrafted methods. The model demonstrates strong generalization ability despite small training datasets due to the fusion strategy and SVM-based decision-making. However, the approach faces difficulties distinguishing subtle expressions and relies heavily on accurate landmark detection. It also lacks evaluation in real-world conditions. Overall, the fusion architecture provides an effective and efficient FER solution that leverages both geometric and appearance representations..</p>