Emotion Detection System Analysis under Partial Occlusion: A Survey

1) Neel Shah (ID:18IT124)
Btech(I.T) 2018-22
Chandubhai S Patel Institute Of Technology
Kundanben Dinshaben Patel Institute of Information Technology
CHARUSAT UNIERSITY
Email id: 18it124@charusat.edu.in

2) Krunal Thakkar (ID:18IT134)
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Chandubhai S Patel Institute Of Technology
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Abstract - Automatic machine-based Facial Expression Analysis (FEA) has gained generous headway in the previous few decades. It has significant applications in the fields of brain science, security, wellbeing, diversion, human-PC associations, and so forth In people feelings assume a critical part for correspondence. They decide how we think, how we impart, how we feel. The dominant part investigations of FEA are impeded from general and tried framework worldwide in controlled climate. Feeling location frameworks chips away at different viewpoints like face, non-verbal communication, voice, body type, skin tone, and so forth Look give some essential data of a human sentiments. Understanding look is an extreme undertaking to decipher relational practices. For the progressing and forthcoming future advancements, look acknowledgment frameworks will assume a significant part in the improvement of human-PC interactions(HCI). For the calculation of feelings in machines, machines need to become familiar with the feelings like people and get them. In this paper, we are looking into the facial feeling identification frameworks and exploration did in this field from various sources accessible all around the world.

Catchphrases: facial feeling recognition Analysis, feeling location framework, fake neural organization, profound learning, AI, man-made reasoning, picture representation, picture handling.

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programming languages like python, matlab, etc. are used in this technology.

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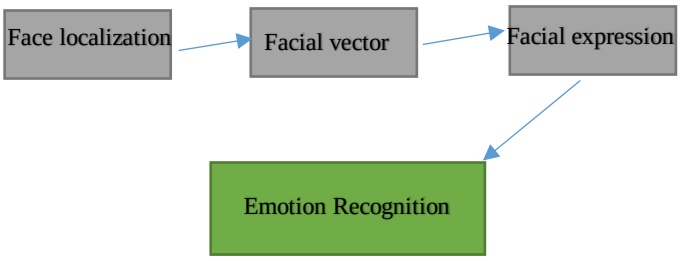

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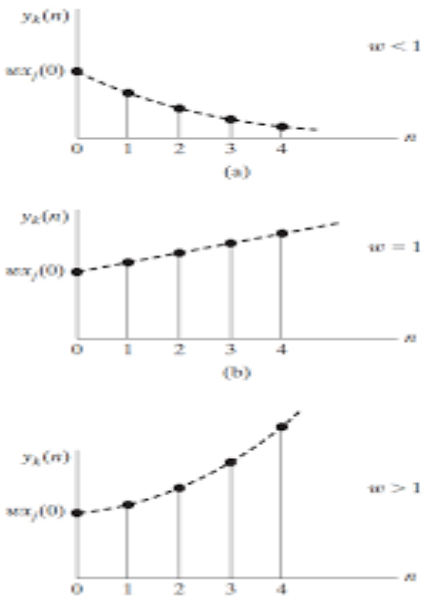

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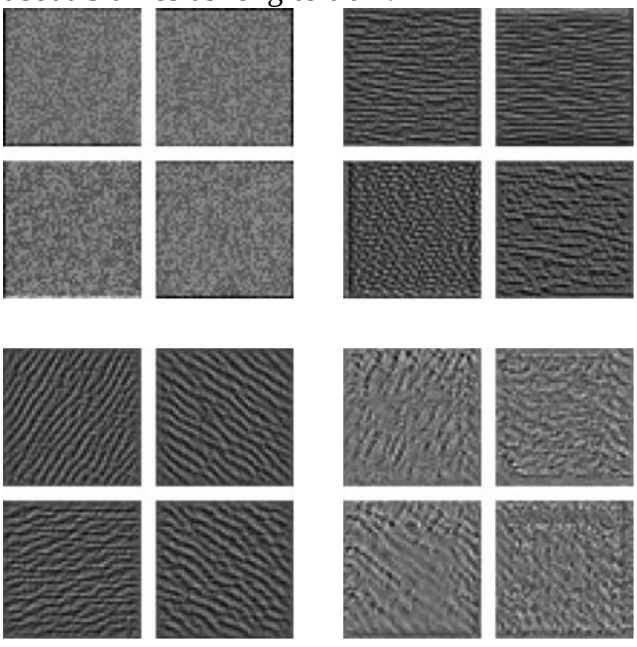

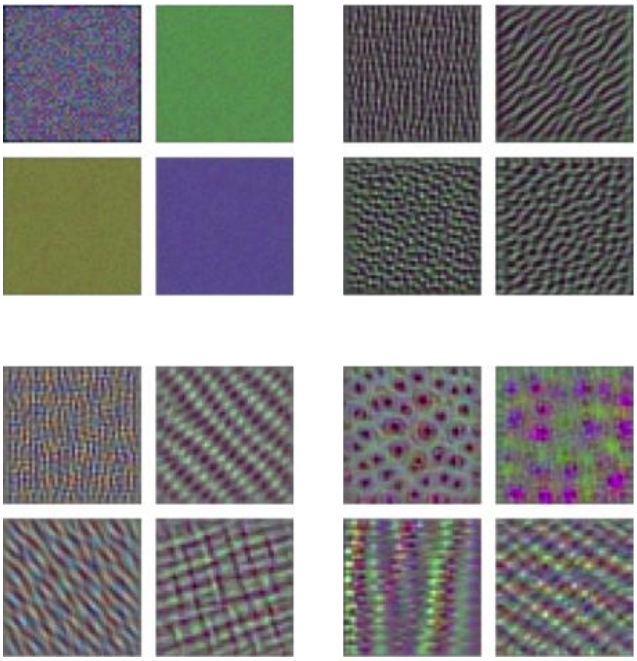

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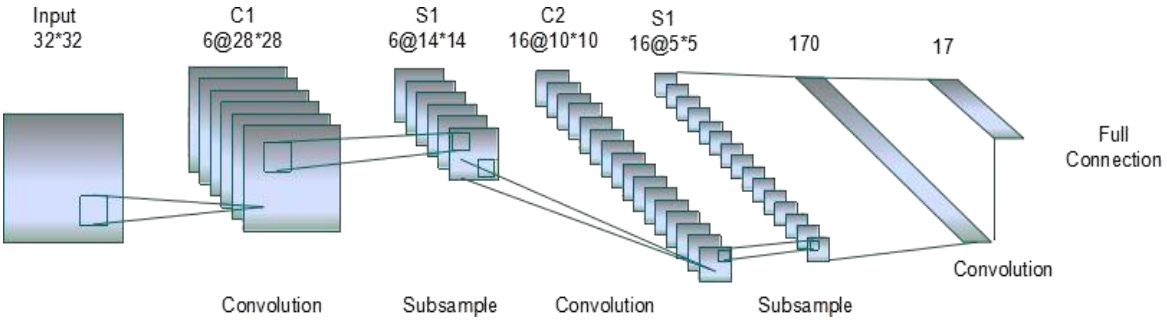

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International Journal of Engineering and Applied Sciences (IJEAS) ISSN: 2394-3661, Volume-3, Issue-2, February 2016

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Evaluation Context

Agnieszka Landowska, Jakub Miler

Gdansk University of Technology, Narutowicza St. 11/12, 80-233, Gdansk, Poland

18IT124,18IT134

by Neel Shah

Submission date: 26-Apr-2021 07:50PM (UTC+0530)
Submission ID: 1570258234
File name: survey_paper_prototype.docx (700.41K)
Word count: 2052
Character count: 12000

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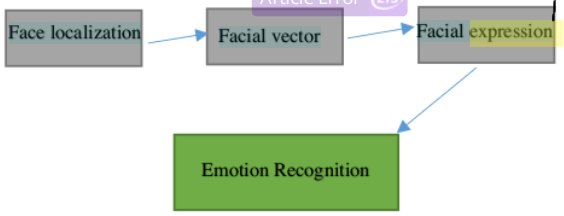

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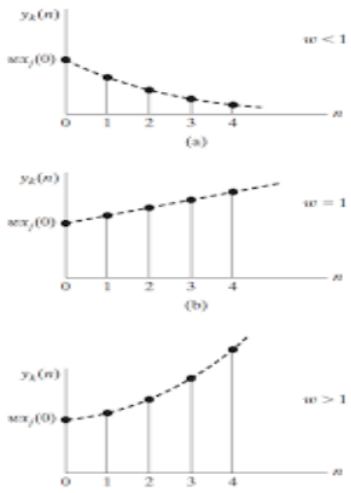

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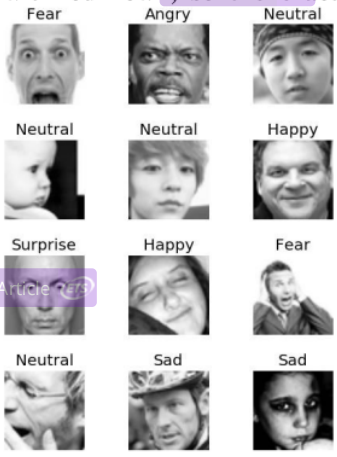

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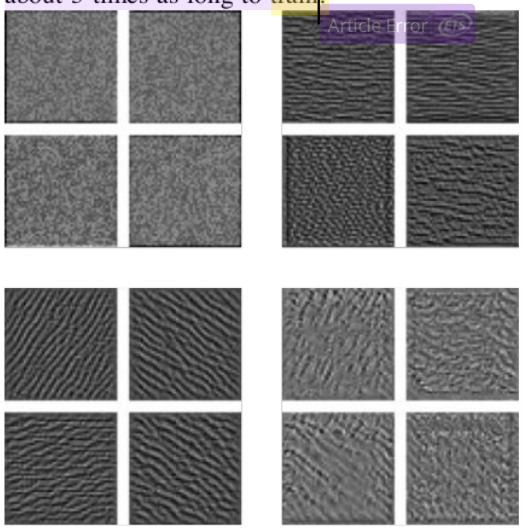

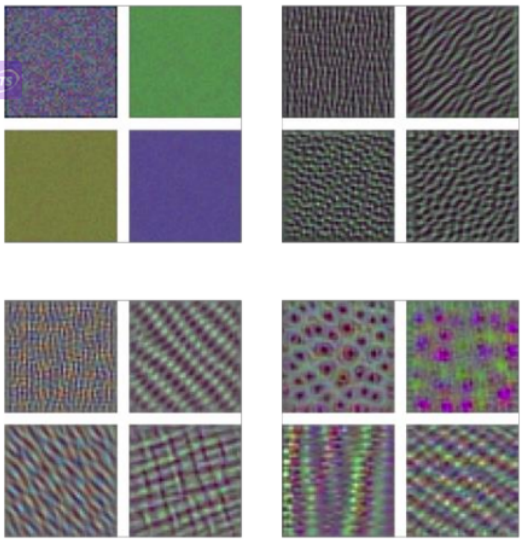

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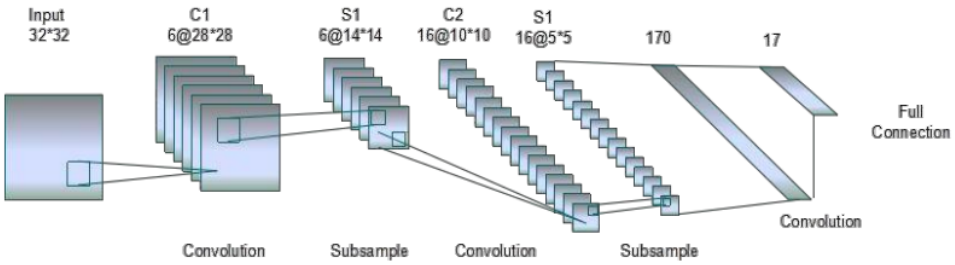

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PAGE 1

Article Error You may need to use an article before this word.

Frag. This sentence may be a fragment or may have incorrect punctuation. Proofread the sentence to be sure that it has correct punctuation and that it has an independent clause with a complete subject and predicate.

Missing "," You may need to place a comma after this word.

Missing "," You may need to place a comma after this word.

P/V You have used the passive voice in this sentence. Depending upon what you wish to emphasize in the sentence, you may want to revise it using the active voice.

Article Error You may need to remove this article.

Article Error You may need to use an article before this word.

Missing "," You may need to place a comma after this word.

Missing "," You may need to place a comma after this word.

Prep. You may be using the wrong preposition.

Article Error You may need to use an article before this word.

Proofread This part of the sentence contains a grammatical error or misspelled word that makes your meaning unclear.

Prep. You may be using the wrong preposition.

Wrong Article You may have used the wrong article or pronoun. Proofread the sentence to make sure that the article or pronoun agrees with the word it describes.

Article Error You may need to remove this article.

PAGE 2

Article Error You may need to use an article before this word.

S/V This subject and verb may not agree. Proofread the sentence to make sure the subject agrees with the verb.

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Sp. This word is misspelled. Use a dictionary or spellchecker when you proofread your work.

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PAGE 5

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PAGE 6

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