

Developing a Facial Landmark Analysis Pipeline for Real Time Scene Authentication to Protect Against Deepfake Attacks

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Introduction: “Deepfake” content refers to visual and auditory media that has been generated by an artificial intelligence (AI). The most recent deepfake technology is capable of generating content that’s indistinguishable from real media, introducing a new class of security threats. Efforts to distinguish deepfake videos from real videos typically involve training a secondary AI algorithm to recognize specific artifacts or physiological cues. However, these countermeasures assume that deepfake generating algorithms won’t continue to evolve and improve. To circumvent a deepfake generation vs detection arms race, we propose a real time scene authentication tool that authenticates an entire scene happening in the real world. This Summer’s project focused on developing a facial landmark analysis pipeline to be used in both the scene and video authentication components of this project.

Method: Using Google Mediapipe’s Facial landmark extraction tool, we can extract and track 478 landmarks on any individual’s face and encode that information into a signal that’s projected onto a scene with imperceptible light. We built a landmark data processing and testing pipeline in Python using the Matplotlib, Scipy, Numpy, and Pandas libraries.

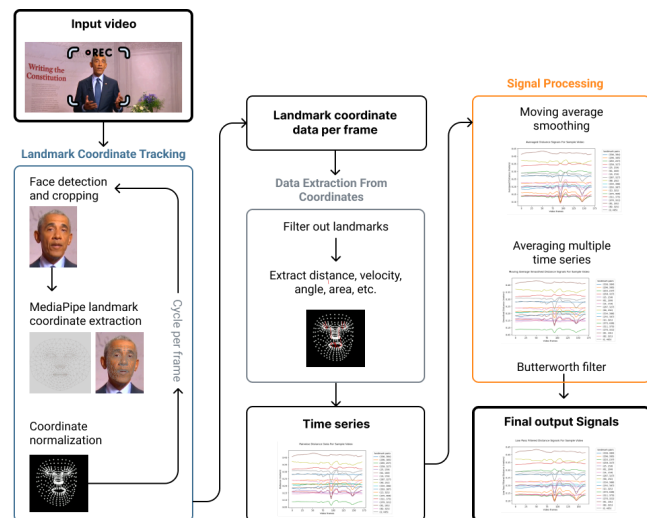


Figure 1. An overview of the pipeline

Results: The main outcome of this Summer is a modular landmark processing and testing pipeline that can be used to determine which landmarks best represent the unique facial movements and identity of a speaker. This pipeline can also be used to determine an optimal sequence for data processing. The current iteration of the pipeline produces the following results:

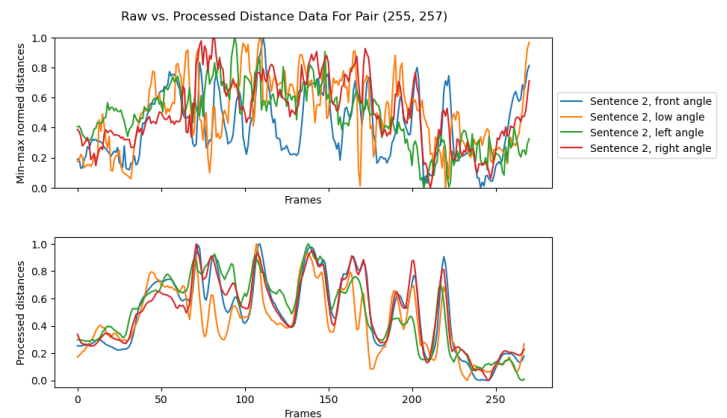


Figure 2. Passing distance signals through the pipeline results in converged signals for videos capturing the same scene.

Future Work: The landmark analysis pipeline needs to be tested on video test data collected from subjects of different age, race, and gender identities. Once more data has been collected, the pipeline can be used to finalize the feature extraction and processing component of the scene authentication project.

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