

# Content

1 USE CASE

2 USE CASE REQUIREMENTS

3 QUANTITATIVE DESIGN REQUIREMENTS

4 SOLUTION APPROACH

5 SYSTEM SPECIFICATION / BLOCK DIAGRAM

6 IMPLEMENTATION PLAN

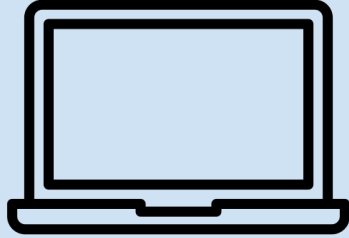
7 TEST, VERIFICATION, AND VALIDATION

8 PROJECT MANAGEMENT

# Use Case

- The problem: Real time instruction is costly and inaccessible, especially for learning casual dances from social media
- Our solution: A virtual, AI powered dance coach that uses your webcam to analyze your moves and provide feedback as you follow along with any dance video
- Tracks and scores dance moves using computer vision tools
- Uses a webcam for motion tracking and 3D modeling
- Provides an accessible and inexpensive learning tool for casual dancers

# Use Case Requirements



Hardware  
Accessibility



Input Video  
Compatibility

# Computer Vision Requirements

<b>Processing speed</b>	Feedback on a 3 minute dance should take no more than 1 minute to generate on consumer grade hardware (Apple M3)
<b>Depth</b>	Full 3-dimensional movement reconstruction from 2D video
<b>Granularity</b>	$\geq 20$ key body points tracked
<b>Accuracy</b>	$\leq 10$ cm for limb position tracking
<b>UDP-Based Networking</b>	CV packet transmission rate of $\geq 20$ packets per second

# Feedback Requirements

<b>Score Accuracy</b>	Per-limb accuracy scores within $\pm 10\%$
<b>Timing Tracking</b>	Timing deviation measured in $\pm 50\text{ms}$ increments
<b>3D Modeled Avatar Representation</b>	Generate corrections for moves with $>20\%$ deviation from reference input video
<b>Improvement Metrics</b>	Track improvement across 5+ key metrics
<b>Progress Reporting</b>	Generate progress reports after every 5 attempts
<b>Performance Benchmarking</b>	Maintain $\geq 30$ FPS in Unity

# Solution Approach

## 1. Motion Capture:

- OpenCV + AI-Driven LandMark Pose Detection

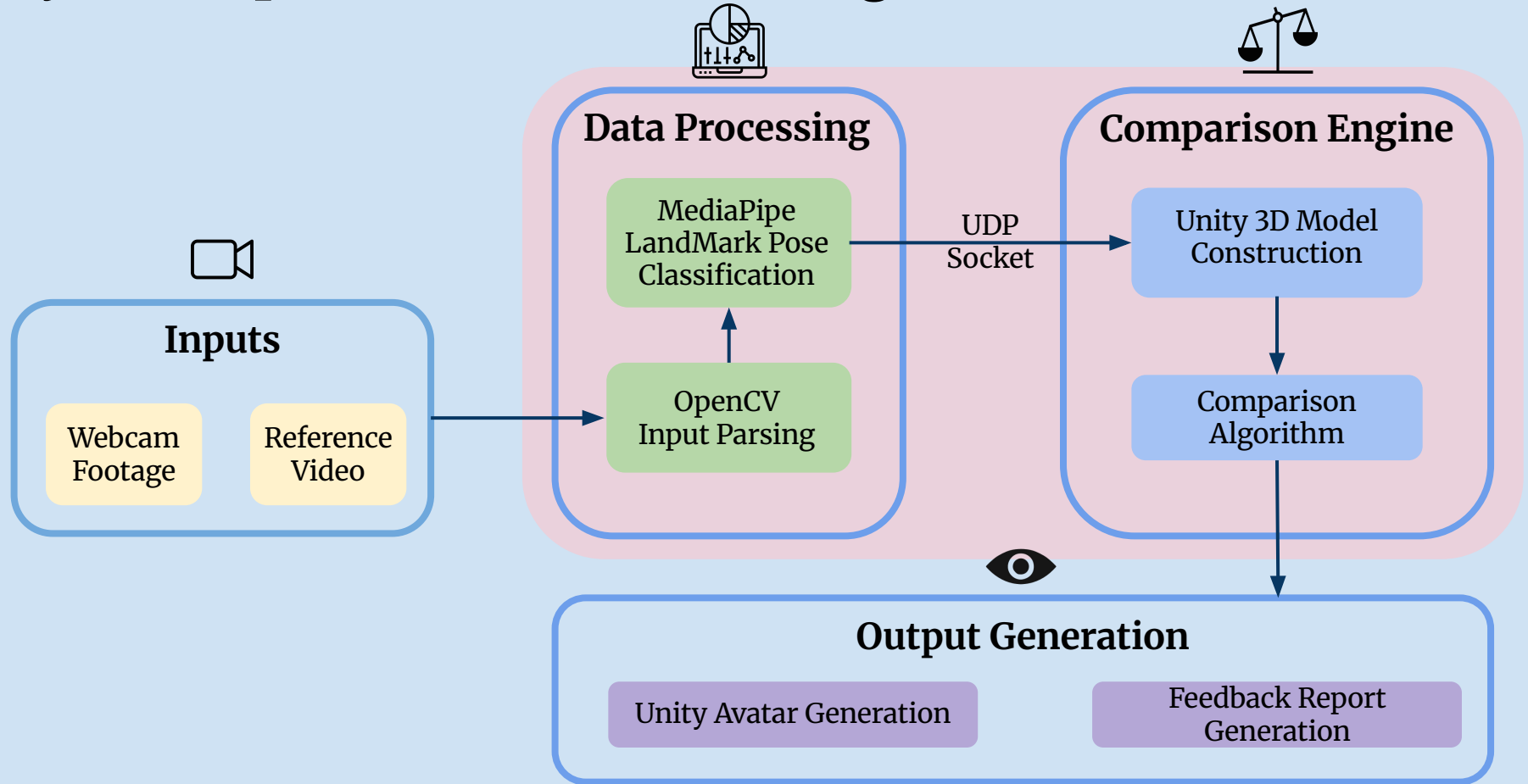
## 2. Reference Comparison:

- Unity based 3D comparison
- Dynamic Time Warping algorithm

## 3. Feedback:

- Joint-specific + Movement pattern heuristics
- 3D Avatar generation

# System Specification/Block Diagram



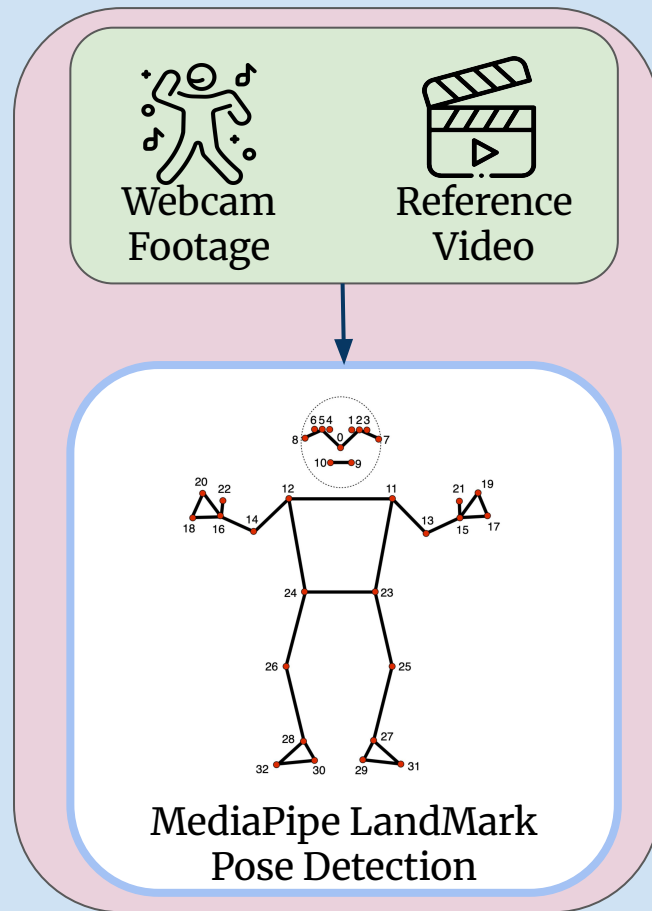
# Implementation Plan - Computer Vision

What:

- OpenCV input processing
- MediaPipe LandMark Pose Detection

Why:

- Accessibility: No need for multiple cameras
- Accuracy: Tried and true libraries





# Implementation Plan - 3D Comparison Engine

What:

- Cosine Similarity + Procrustes Analysis

Why:

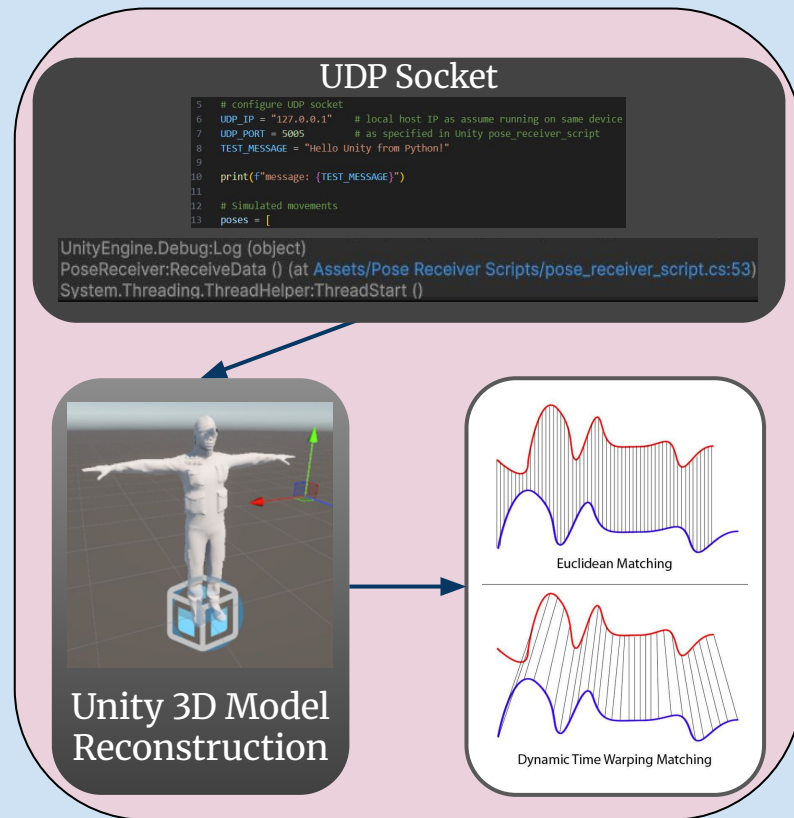
- Normalizes spatial variance in pose keypoints, enabling rotation/scale-invariant comparison

What:

- Dynamic Time Warping Algorithm

Why:

- Optimizes temporal alignment between movement sequences, accommodating non-linear timing variations and preserving sequential pose correspondences



# Test, Verification and Validation

## Computer Vision:

<b>Processing speed</b>	Input ~3 minute dances on Apple M3 laptop
<b>Depth</b>	Input complex dances that heavily involve 3D movement
<b>Granularity</b>	Count # of body points tracked consistently for the entire duration of the video input with at least 5 different human targets
<b>Accuracy</b>	Cross compare processed data with input for $\leq 10$ cm discrepancy
<b>UDP-Based Networking</b>	Utilize Python packet sending and timer to ensure sufficient throughput

# Test, Verification and Validation

## Feedback:

<b>Score Accuracy</b>	Compare output scores with at 3 preset dance videos to remain within $\pm 10\%$ deviation from the reference movements
<b>3D Modeled Avatar Representation</b>	Test dance sequence with known deviation quantities to verify system generates movement corrections for $>20\%$ pose deviations
<b>Progress Reporting</b>	Ensure that a detailed progress report is generated after every 5 attempts, summarizing performance trends
<b>Performance Benchmarking</b>	Measure Unity frame rate ( $\geq 30$ FPS) while providing 3D avatar feedback

