

Sensor Fusion System for Improving Motion Amount Quantification within CV-Enabled Worker Analysis



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Research Question

How can computer-vision-enabled motion amount quantification be improved **without** changing the computer-vision model in a way that is **simple**, **inexpensive** and **effective**?

Approach Brainstorming

- Optimize Data Preprocessing and Feature Engineering
 - Ensuring data diversity and consistency is a **user effort**
 - Optimizations should happen during the **training process**
- Using Ensemble Methods
 - Combining multiple outputs is **computationally taxing**
 - Aggregation techniques are both **simple** and **reliable**
- Incorporate Richer Contextual Information
 - Combining multiple outputs is **computationally taxing**
 - May **perform better than simple aggregation** due to different types of data
 - More **expensive** due to requiring different sensor types
- Perform Data Post-Processing
 - Additional filtering and smoothing will help **reduce noise** and **improve stability** of motion quantification

Chosen Approach + Reasoning

Elimination:

- #1 changes the model and requires user effort ▶ **Rejected**
- #2 is costly but simple and reliable ▶ **Accepted**
- #3 is very costly but performs well ▶ **Accepted**
- #4 has no downsides except developer investment ▶ **Accepted**

Choosing between #2 and #3:

Considering the context, worker analysis for general motion amount quantification: **Soft real-time system**.

If the system performs **reasonably well**, cost is of **more importance** to the target user than an improvement to output accuracy. Thus, choose #2.

Final Choice:

Use Ensemble Method (homogeneous sensor fusion) with Data Post-Processing (filtering)

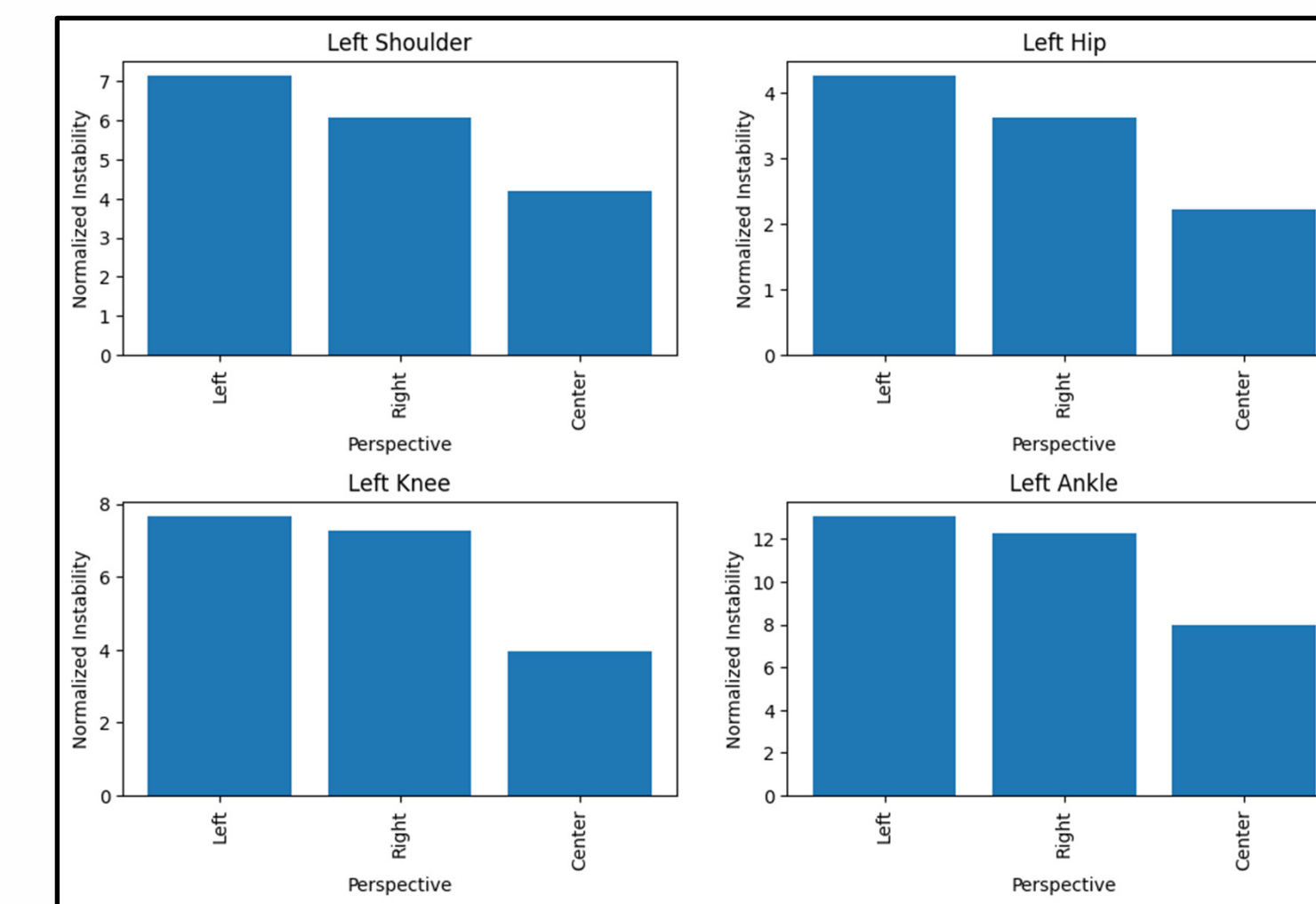
Current Progress



```
jointn_most_stable_perspective()
    ↓
joint1_angle
    ...
jointn_angle
    ↓
calculate_output()
```

What the output signifies and how it is calculated is decided per implementation and is left modular for different use cases.

- State (Dangerous, Safe)
- Efficiency (100% - 0%)
- Mood (Happy, Sad)
- Instability itself (see below)



`instability` represents the cumulative Euclidean distance traveled by a specified joint at a specified frame. This way of measuring accuracy was chosen for (1) its **simplicity** and (2) because **erratic movement of landmarks** in the video feeds were the most common symptom of bad output.

Project Objective

Increasing the **accuracy** of computer-vision-enabled motion amount quantification in a **simple** and **intuitive** way, with **little to no** user investment.

Future Directions

- Optimizing the performance of the sensor fusion system so lower-end devices can reliably perform computations in real-time
- Improving the instability metric to something better than Euclidean distance
- Creating other presets for calculate_output() other than instability and state, with accompanying control system
- Exploring model architecture tuning and other techniques that are not in the scope of this project

Acknowledgements

The pose estimation model I used to process landmarks is a MediaPipe solution that you can find at this link - <https://developers.google.com/mediapipe/solutions>

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