

# Final Term Project Presentation

## Skeleton-based Gait Recognition via Robust Frame-level Matching

Md Bakhtiar Hasan

STUDENT ID: 181041013

Department of Computer Science and Engineering  
Islamic University of Technology

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- **Gait**  
How a person walks
- **Gait Recognition**  
Biometric identification method
- **Skeleton-Based**  
Skeleton joints extracted by 3D motion capture tool
- **Frame-Level Matching**  
Matching based on the quality of the frame

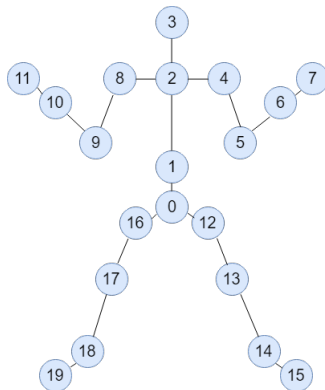


Figure: Extracted Skeleton Joints



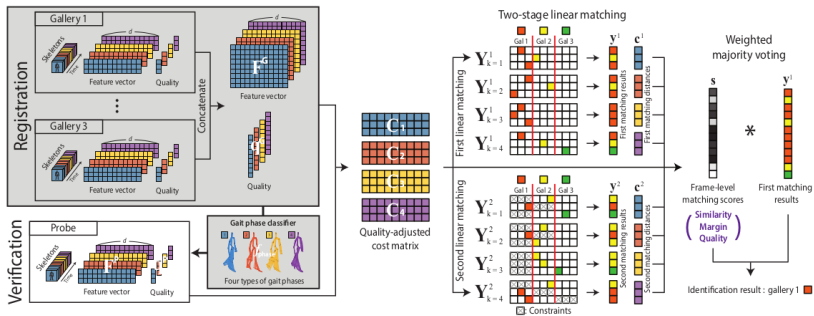
- Development of view and scale-invariant feature descriptor
- Effective representation of human gait
- Ensuring low computational cost
- Minimize the influence of noise and other covariate conditions



- Appearance-based methods [1], [2]
  - Silhouette information obtained from background subtraction
  - Average aligned silhouettes through statistical analysis - also known as Gait Energy Image
  - Low computational cost
  - Sensitive to changes in observation view
- Model-based methods[3], [4], [5], [6], [7]
  - Analyze the kinematics of human body parts in 3D space
  - Use sensors or motion capturing devices for constructing 3D skeleton
  - Inaccurate skeleton estimation as noisy frames and characteristic frames are equally treated

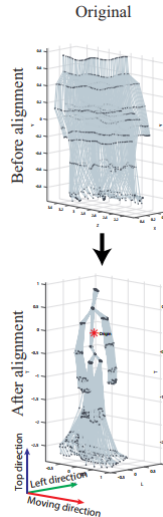


# Overall Framework



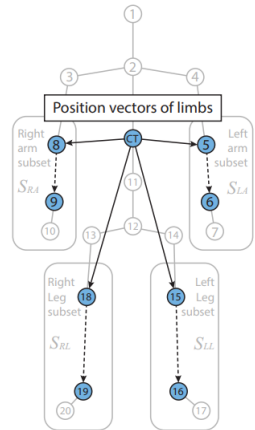
# Centroid-based Skeleton Alignment

- Translation vector to move the centers of all skeletons to the origin
  - using centroid of joint position of torso
- Scale vector to make all skeletons equal in size
  - using centroids of the joints of upper torso and lower torso
- Rotation matrix to rotate all skeletons in the same direction
  - using position difference of the centroid to detect moving direction
  - position of centroid taken in variable time interval



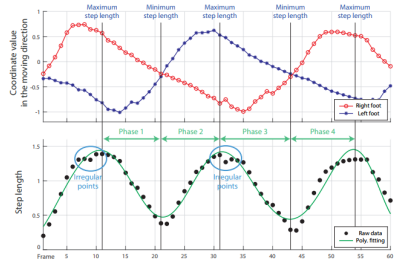
# Spatial Modeling of Gait Pattern

- Combination of position vectors from centroid
  - Left Elbow
  - Left Knee
  - Right Elbow
  - Right Knee
- Other vectors
  - Left Elbow  $\rightarrow$  Left Wrist
  - Left Knee  $\rightarrow$  Left Ankle
  - Right Elbow  $\rightarrow$  Right Wrist
  - Right Knee  $\rightarrow$  Right Ankle
- Contains information about both distance and angles
- Invariant to view and scale due to skeleton alignment
- 8 position vectors to form a 24-dimensional feature vector



# Temporal Segmentation of Gait Cycle

- Divide recurrent gait cycle into 4 gait phases based on the movement of feet
- Estimate label of each gait phase based on which foot is in front
- Polynomial fitting to the raw step length to reduce irregularities in step length measurement
- Use Random Forest classifier to estimate labels





# Two-stage Linear Matching

## Preprocessing

- Pairwise distance between two vectors - Manhattan distance
- Measure quality of skeleton based on symmetry of human body using arm and leg symmetry
- The more assymmetric the length of the arm or leg, the lower the skeleton quality
- Create cost matrix using distance of test frame and stored frame, and quality
- Compute cost matrix for phases of stored frame that are identical to phases of test frame to reduce computational cost



# Two-stage Linear Matching

## Stage 1

- Linear assignment problem - Combinatorial optimization to find minimum weight matching given a cost matrix  
Solution: Hungarian Algorithm
- Solve linear assignment problem with cost matrix to find the person having the most similar pattern from stored frames

## Stage 2

- Again solve linear assignment problem with second-most similar pattern
- To prevent matching with same person give infinite cost penalty to the first matched person



# Weighted Majority Voting

- Using quality of the input frame, assign weight to each of the matched phases
- Calculate similarity of each pattern using inverse of distance from stored pattern
- Calculate margin between two matched patterns using ratio of their distance from the matched frame
- Calculate frame-level matching score using the combination of these three
- Vote the user with the most matching score



# Implementation

## Referenced Paper

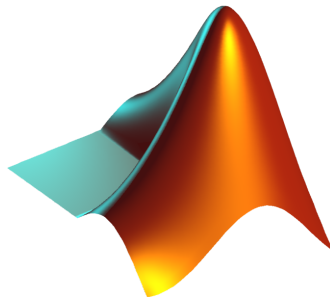
Choi, S., Kim, J., Kim, W., & Kim, C. (2019). Skeleton-based Gait Recognition via Robust Frame-level Matching. *IEEE Transactions on Information Forensics and Security*.

## Tools Used

- MATLAB

## Dataset

- UPCVgait
- CILgait
  - CIL-S (Stops Walking)
  - CIL-SC (S+Calling Pose)
  - CIL-SV (S+Looking at phone)
- UPCVgaitK2 (Not Freely Available)
- SDUgait (Not Freely Available)



# Result Analysis

## Accuracy

Method	UPCV1	CIL-S	CIL-SC	CIL-SV
Implementation	99.07	87.50	75.00	83.33
Ball et al. [3]	58.09	8.33	4.17	0.00
Preis et al. [4]	81.58	20.83	20.83	16.67
Kastaniotis et al. [5]	70.22	33.33	16.67	33.33
Ahmed et al.[6]	82.13	20.83	12.50	12.50
Kastaniotis et al. [7]	94.11	58.33	41.67	54.17

## Runtime on UPCV1 dataset

Method	Runtime (seconds)
Implementation	3.05
Ahmed et al. [6]	8.07
Kastaniotis et al. [7]	61.58



- Reducing frame-by-frame comparison by finding representative patterns
- Use other image based 3D pose estimation techniques instead of Kinect



# References

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- ④ Preis, J., Kessel, M., Werner, M., & Linnhoff-Popien, C. (2012, June). Gait recognition with kinect. In *1st international workshop on kinect in pervasive computing* (pp. 1-4). New Castle, UK.
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- ⑥ Ahmed, F., Paul, P. P., & Gavrilova, M. L. (2015). DTW-based kernel and rank-level fusion for 3D gait recognition using Kinect. *The Visual Computer*, 31(6-8), 915-924.
- ⑦ Kastaniotis, D., Theodorakopoulos, I., Theoharatos, C., Economou, G., & Fotopoulos, S. (2015). A framework for gait-based recognition using Kinect. *Pattern Recognition Letters*, 68, 327-335.



# THANK YOU

