You Are How You Walk: Uncooperative MoCap Gait Identification for Video Surveillance with Incomplete and Noisy Data

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Identification Pipeline



Phase I – Acquiring Motion Capture Data

Motion capture (MoCap) technology provides Frame: 0 video clips of moving individuals containing an overall structure of the human body and estimated **3D joint coordinates**.

MoCap data can be collected online by a readily available system of multiple cameras (Vicon) or by a depth camera (Microsoft Kinect).

To visualize MoCap data, a stick figure that represents the human skeleton can be recovered from joint spatial coordinates in time.



Phase II – Detecting Gait Cycles



People spotted in our tracking space do not Clean gait cycles need to be first filtered out. walk all the time; on the contrary, they perform various activities.

Identifying people from gait requires video segments where they are actually walking.

There are methods for detecting gait cycles directly as well as **action recognition** methods that need a demonstrative example of a gait cycle to query general motion sequences.



Phase III – Extracting Gait Features

Let a model of the human body have J joints and all N_L learning samples of C_L walkers be linearly normalized to their average length T. Labeled learning data in a **sample space** have the form $\mathcal{G}_L = \{(\mathbf{g}_n, \ell_n)\}_{n=1}^{N_L}$ where

 $\mathbf{g}_{n} = \left[\left[\gamma_{1} \left(1 \right) \cdots \gamma_{J} \left(1 \right) \right] \cdots \left[\gamma_{1} \left(T \right) \cdots \gamma_{J} \left(T \right) \right] \right]^{\mathsf{T}}$

is a gait sample (one gait cycle) in which $\gamma_i(t) \in \mathbb{R}^3$ are 3D spatial coordinates of joint $j \in \{1, \ldots, J\}$ at time $t \in \{1, \ldots, T\}$ normalized to the person's position and walk direction.



Each learning sample falls into one of the learning **identity classes** $\{\mathcal{I}_c\}_{c=1}^{C_L}$ labeled by ℓ_n . **Feature extraction** is set by matrix $\mathbf{\Phi} \in \mathbb{R}^{D \times \widehat{D}}$ from *D*-dimensional sample space $\mathcal{G} = \{\mathbf{g}_n\}_{n=1}^N$ to \widehat{D} -dimensional **feature space** $\widehat{\mathcal{G}} = {\{\widehat{\mathbf{g}}_n\}}_{n=1}^N$ with D < D.

A given gait sample \mathbf{g}_n can be transformed into gait template by $\widehat{\mathbf{g}}_n = \mathbf{\Phi}^\top \mathbf{g}_n$.

Templates are compared by the Mahalanobis distance.

Evaluated **methods**:

- 8 hand-designed geometric feature sets
- Features learned by **MMC**
- Features learned by **PCA+LDA**

Evaluated **metrics**:

- Correct classification rate
- Discriminativeness
- Robustness to noisy and incomplete data
- Clusterability
- Scalability



Phase IV – Identifying Walkers



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