mSIMPAD: Repetitive Activity Detection

What?

 Identify if a person performs repetitive activity using data collected from mobile devices.



Typical human activity recognition (HAR) methods:



Possible Solution II: Repetitive Activity Detection

 Determine if the input data contains repetitive patterns with Autocorrelation / Frequency Domain Analysis



Proposed Solution

- Find if there contains any repetitive patterns in the time series
 - Generality unlimited type of repetitive activities
 - Robustness handle variable shape, length, and interval



Problem Formulation

Definitions



• Given $T, L = (l_1, ..., l_{|L|})$, and m, identify all subsequences in T that contains successive similar patterns (SSP)

Range-Constrained Matrix Profile (RCMP)

- **RCMP** is a variation of **Matrix Profile** [Yeh'16], a data structure that annotate a time series.
 - A value of RCMP *MP_l* at index *i* is the **distance** between *T_{i,l}* to its **nearest neighbor** *T_{j,l}* within the **user-defined range** *m*.



SSP with Different Lengths



What can we learn from this?

- MP containing SSP has relatively lower value.
- Regions that lower than θ are called **valleys**.
- Valley with larger area depth is a better fit to the pattern.

Choosing The Best Set of Valleys

- Finding the best match of valleys = selecting valleys that maximize the total sum
- Let $V = \{v_1, v_2, \dots, v_{|V|}\}$ be the set of valleys found in *MP* at all *l*.
- $idx(v_i)$ denotes the corresponding indices of v_i .

$$\max_{V_{opt} \subseteq V} \sum_{v_i \in V_{opt}} v_i$$

s.t. $\forall (v_i, v_j) \in V_{opt}: idx(v_i) \cap idx(v_j) = \emptyset$

 Reduce to a maximum-weight independent set problem, and solved with a branch-and-bound approach

Example of Repetitive Activity Detection



- **Fewer** false positives
- More accurate and coherent detected regions

Evaluation on Detection Performance

Dataset	Algorithm	Accuracy (%)	Precision (%)	Recall (%)	F-Score (%)
НАРТ	NASC	94.10 ± 3.49	91.26 ± 3.05	91.56 ± 10.88	91.07 ± 6.57
	STFT	94.03 ± 2.37	91.30 ± 3.35	91.73 ± 5.44	91.41 ± 3.34
	SIMPAD	$\textbf{96.44} \pm \textbf{2.26}$	95.63 ± 4.33	94.50 ± 3.89	94.96 ± 2.98
	mSIMPAD	96.16 ± 2.60	94.35 ± 5.36	95.10 ± 3.45	94.62 ± 3.44
PAMAP2	NASC	81.70 ± 12.32	$\textbf{99.39} \pm \textbf{0.85}$	66.47 ± 21.95	77.12 ± 22.26
	STFT	78.79 ± 9.12	99.31 ± 0.90	62.45 ± 6.75	76.50 ± 4.76
	SIMPAD	84.11 ± 5.59	99.12 ± 1.14	71.24 ± 5.66	82.78 ± 3.74
	mSIMPAD	84.62 ± 5.65	98.28 ± 1.78	$\textbf{72.90} \pm \textbf{5.58}$	$\textbf{83.59} \pm \textbf{3.62}$

Outperforms baseline methods

 Robust to noisy and poor-quality data

Performance on HAPT [Anguita'13] and PAMAP2 [Reiss'12] where the values given as mean \pm SD.



Evaluation on Running Time







- Compute large amount of data in minutes
- Scales linearly to the input size
 - support real-time applications