Detecting fatigue in sports data via time-uniform martingale bounds

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1. Introduction

- Data is obtained from runners over a long race via (inertial)motion sensors attached to the body or force plates (embedded within shoes or treadmill).
- The goal is fatigue detection for data collected over time by studying change in the stream of the sensor data. This means segmenting data into <u>rest</u> and fatigue states.
- A centered martingale is constructed from the data which serves as the test statisic, with critical thresholds given by linear and LIL (law of iterated logarithm) bounds over time.
 Results: Segmentation of data into rest, fatigue states. Comparison of average stride patterns from different states leads to region of interest (ROI) identification in functional data.
 Applications areas would be any form of gait analysis in sports and medicine and change-point detection problems.

2. Data, test problem, critical threshold

- We use the functional signal plus noise model where each stride Y_i , i = 1, ..., n is represented as $Y_i(t) = \mu_s(t) + \epsilon_i(t)$, with $\mu_s(t)$ as an average stride pattern for $s \in \{\text{rest, fatigue}\}$ and ϵ_i as gaussian error.
- Each of the functional curves are reduced to a single point by means of L²-error: Δ_{i,s} := ||Y_i(t) − μ_s(t)||²₂ as in (3).
 For the null H_i : Δ_{i,s} ^{i.i.d.} F, we define test problem:

 $\mathcal{H}_t = \cap_{i=1}^t H_i.$

- With δ as level of test \mathcal{H}_t , consider centered martingale M_t given by, $M_t = \sum_{i=1}^t \mathbb{1}\{\Delta_{i,s} > \gamma\} t\alpha$, where γ is local critical threshold at level- α .
- M_t (shown in (4)) serves as test statistic over time $t \in \mathbb{N}$.

Mathematical analysis of sports data: A complete cycle



Figure 1: (top left) Runner, [wikipedia]; (1) Hip and Knee angles; (2) one stride from knee angles; (3) L^2 -error $(\Delta_{i,s})$; (4) Point data over time and test statistic $M_t(5)$ (left) Linear bounds $L_j(t)$ and (right) finite-LIL bound; (6) Upcrossing of M_t over Γ_t (7) ROI in knee stride and feedback

3. Critical threshold for martingale

Given $\alpha, \delta \in (0, 1), p \in \mathbb{N}, t \in \mathbb{N}$ and $\tau = 1064.3 \log(\frac{1}{\delta})$, we show that for Γ_t (threshold over time t) given by,

$$\int L_j(t), \qquad j = 1, \dots, p; \text{ for } t < \tau$$

4. ROI identification and feedback

- Upcrossing of martingale over the Γ_t (in (6)) bound captures change in underlying distribution of point data in (3) while maintaining overall type-I error δ .
- Test procedure may be performed for online, sequentially arriving functional or point data and results in segmented data

 $\prod_{t=1}^{T} \lim_{t \to \infty} \left\{ L_j(t), \sqrt{1.2t \left(2\log\log t + \log \frac{1.15}{\delta} \right)} \land 1 \land 0.04t \right\}, t \ge \tau$

it holds for the null \mathcal{H}_t with (global-) level δ that,

 $\mathbb{P}_{\mathcal{H}_t}\left(\forall t \in : M_t > \Gamma_t\right) \leq \delta,$ where $L_j(t) \coloneqq \frac{c_j}{\alpha(1-\alpha)t_{0j}}t + c_j$ and c_j, t_{0j} are parameters of L_j . [1, 2] from various phases (or underlying distributions).

- For functional data, detecting time point of change allows for detecting further region of interest (ROI) within functions from different segments.
- Results (for knee angle data): ROI (in figure (7)) indicate increased knee stiffness which can lead to injuries due to reduced shock attenuation (see [3]) and therefore appropriate feedback has to be given to runner for improved training.



References

- [1] Steven R et. al. Howard. Time-uniform chernoff bounds via nonnegative supermartingales. *Probability Surveys.*
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- [3] Jasper et. al. Reenalda. Kinematics and shock attenuation during a prolonged run on the athletic track as measured with inertial magnetic measurement units. *Gait & posture*, 68:155–160.