## A. Supplementary Material | Algorithm for Identification of IFD

## Study Conditions

$$\begin{split} EMG_{param} &= [\text{EMG parameters of Table 1}] \\ HRV_{param} &= [\text{HRV parameters of Table 1}] \\ WS_{EMG} &= [5, 10, 15, 20, 25] \text{ muscular activation periods} \\ WS_{HRV} &= [30, 40, 50, 60, 70, 80, 90, 100, 110, 120] \text{ s} \\ \#Participants_{EMG} &= 14 \\ \#Participants_{HRV} &= 11 \end{split}$$

Study Conditions Applied to the Algorithm  $\{signal\} = EMG \text{ or } HRV$ 

## Generic Algorithm

for each  $P_k$  in  $\{signal\}_{param}$  do  $WS_{best} = None$  $TS_{best} = None$  $CV_{best} = \infty$  $\begin{array}{l} \textbf{foreach } WS_z \textbf{ in } WS_{\{signal\}} \textbf{ do} \\ \mid \quad TS = [0, 10, 25, 50, 75, 90]\% \text{ of } WS_z \end{array}$ for each  $TS_y$  in TS do  $m_{array} = []$  $\sigma_{array}^2 = []$ for n = 1 to  $#Participants_{\{signal\}}$  do if  $CV[WS_z; TS_y] < CV_{best}$  then 1) Generation of the evolution time series of  $P_k$  for a sliding window configuration of window size  $WS_z$  and time-step  $TS_y$ 2) Determination of the regression curve that best fits the time series generated in 1 **3)** Storage of slope  $m_n$  inside  $m_{array}$ 4) Storage of variance  $\sigma_n^2$  inside  $\sigma_{array}^2$ end  $\mathbf{end}$ 1) Determination of  $m_{comb}$  and  $\sigma^2_{comb}$ , accordingly to the mathematical formalism defined in equations 4 and 5  $\,$ **2)** Determination of  $CV[WS_z; TS_y]$  (equation 8) if  $CV[WS_z; TS_y] < CV_{best}$  then  $WS_{best} = WS_z$  $TS_{best} = TS_y$  $CV_{best} = CV[WS_z; TS_y]$  $\mathbf{end}$  $\mathbf{end}$  $\mathbf{end}$ For the combination  $[WS_{best}; TS_{best}]$ , application of criteria  $\mathcal{C}$  (section 2.3.2) in order to verify if  $P_k$  is and Individual Fatigue Descriptor (IFD)  $\mathbf{end}$