

Indoor Environment Control Method for Improving Well-Being using Human Thermal Stress Estimated by Yuragi Learning



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Background

- Individual's good work-life balance are being recommended^[1]
 - Many changes in work style can be burdensome for workers^[2]
- There is still room to consider ways of working
 - Increasing productivity while taking into account personal well-being^[3]
- Biometric information can estimate stress state^[6]
 - Using skin potential activity, skin temperature, and pulse wave



[1] "Abenomics: For future growth, for future generations, and for a future Japan (Online available).
[2] D. E. Guest, "Human resource management and employee well-being: Towards a new analytic framework," Human resource management journal, 2017.
[3] A. Sakuraya, et al., "Job crafting, work engagement, and psychological distress among Japanese employees: A cross-sectional study," bioPsychosocial medicine, 2017.
[6] P. Schmidt, et al., "Introducing wesad, a multimodal dataset for wearable stress and affect detection," in Proc. ACM ICMI, 2018.

Purpose and Problems

- Purpose**
 - Estimating the stress state of each individual from human biological information
 - Controlling the indoor environment based on human stress
 - Individual can stay comfortably
- Problems**
 - The biological information from wearable sensors contains noise
 - Stress affects biometric information differently for each individual
 - Using **Yuragi learning** to solve these problems

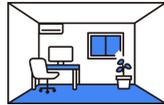


Image of comfortable room

Even in the same indoor environment

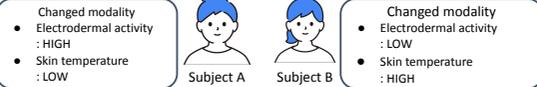
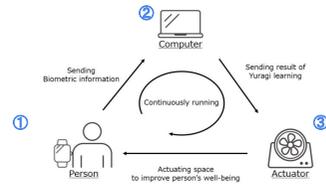


Image of Stress affects biometric information differently person to person

Proposal

- Realizing well-being space by 3 steps
 - Sense various biometric information using biometric sensors
 - Estimate a person's psychological state from the biometric information
 - Intervene indoor space by controlling actuators
 - To relieve the stress state
- We implement this system and conduct experiments



Overview of system

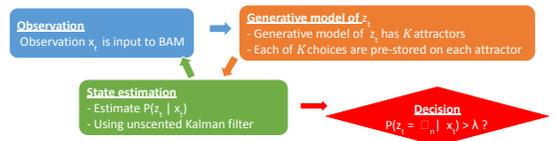
Wearable Stress and Affect Detection (WESAD)^[6]

- Biometric information can estimate human emotional state
 - Various biometric information from 15 subjects
 - Blood volume pulse (BVP), Electrocardiogram (ECG), Electrodermal activity (EDA), Electromyogram (EMG), Respiration (RESP), Skin temperature (TMP), and 3-axis accelerometer (ACC)
 - Their self-reported is used as emotional-state label
 - Neutral, Stress, Amusement
- They achieved high accuracy in estimating psychological states
 - Using machine learning methods on these data
 - Decision trees, Random forests, AdaBoost, Linear discriminant analysis, K-nearest neighbor methods
- Our research use as reference this study
 - Our research uses various biometric information too
 - Our research uses **Yuragi learning**^[7] for estimating stress state

[7] M. Murata and K. Leibnitz, Fluctuation-Induced Network Control and Learning: Applying the Yuragi Principle of Brain and Biological Systems. Springer, 2021.

Bayesian attractor model (BAM)

- BAM models the decision process that human brain makes
 - BAM decides which of choices the observed information corresponds to
 - This decision-making can be made even from noisy information



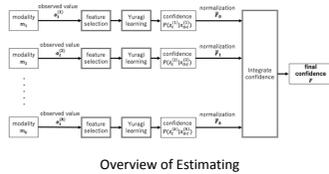
Overview of BAM

- Decisions are made according to confidence level
 - Confidence level means that how z_t is close to one of the attractors z_k
 - Adopt Attractor whose confidence level exceeds the threshold λ
 - This allows decisions to be made with fewer errors for noisy observations

Multimodal recognition based on Yuragi learning

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- Using BAM for decision making machine learning methods
 - Our research group has succeeded in various applications^[7]
 - Can be run on relatively small systems
- Composed of frameworks that use BAM
 - Input observed information (preprocessing)
 - Set attractor (training)
 - Make decision
- In this study, integrating multiple biometric outputs



Overview of Estimating

Experimental setup

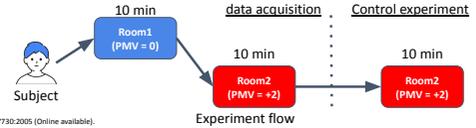
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- Two rooms were set up as experimental environments
 - Indoor environment is set based on PMV^[14]
 - Comfortable room was set as PMV = 0 (20°C, 40% humidity)
 - Uncomfortable room was set as PMV = +2 (30°C, 70% humidity)



Experiment Room

- Experimental steps
 - Acquisition of tain data (10min * 2)
 - Actuator-control experiment (10min)
 - Estimated subject's state is checked every 10 seconds
 - Actuator changes a room environment according to this state



[14] ISO 7730:2005 (Online available).

Devices

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- Empatica E4 wristband^[9]
 - It used for acquiring ACC, BVP, EDA, IBI, TMP
- Laptop computer
 - Installs a streaming server which receives biometric information
 - Connected to the E4 wristband via Bluetooth
 - Conducts Yuragi learning
 - Sends control commands to an actuator by the estimated result
- Daikin assisted circulator^[10]
 - The circulator is connected to a Raspberry Pi 3 model B (PI3)
 - its operation can be changed by serial commands sent from the PI3
 - Blowing wind when stress state is estimated



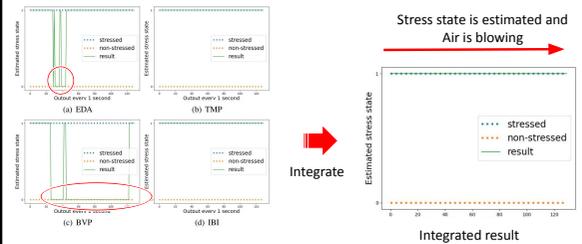
E4 wristband and assisted circulator

[9] "E4 wristband," <https://www.empatica.com/en-int/research/e4/>. (Accessed on Nov. 1, 2023).
 [10] "Daikin assisted circulator," <https://baunch.jp/dalists.com/en/feedback/detail/>. (Accessed on Nov. 1, 2023).

Result

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- Stress state are estimated and control execution is confirmed
 - Green line represents the estimated state at each time step
 - Improved accuracy by combining multiple biometric data
 - In this experiment, the control always runs



Result of Estimation

Conclusion

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- We proposed a Yuragi Learning based stress estimation method
 - Integrating multiple types of biometric information
 - Experiments have shown that the system can operate correctly
- Future works
 - How to control the actuators to completely remove stress
 - Pursuit of adequacy of actuating methods
 - How to estimate and control them for different people at the same time
 - Another experiment scenario
 - Control PMV by changing exercise
 - In this research we control PMV by temperature and humidity

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- [8] S. Bitzer, J. Bruineberg, and S. J. Kiebel, "A Bayesian attractor model for perceptual decision making," *PLoS Computational Biology*, vol. 11, no. 8, p. e1004442, 2015.
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