
Autonomy is the Key: From Smart Towards Intelligent Textiles

Olivia Ojuoye

EEE Research Group
 Electronics and Computer Science
 University of Southampton
 SO17 1BJ
 United Kingdom
 oo2g12@ecs.soton.ac.uk

Russel Torah

EEE Research Group
 Electronics and Computer Science
 University of Southampton
 SO17 1BJ
 United Kingdom
 rnt@ecs.soton.ac.uk

Steve Beeby

EEE Research Group
 Electronics and Computer Science
 University of Southampton
 SO17 1BJ
 United Kingdom
 spb@ecs.soton.ac.uk

Adriana Wilde

WAIS Research Group
 Electronics and Computer Science
 University of Southampton
 SO17 1BJ
 United Kingdom
 agw106@ecs.soton.ac.uk

Abstract

Electronic textiles become smart by embedding circuits and sensors which offer some passive or active capabilities. Smart textiles become intelligent due to their computational abilities allowing awareness of their environment, extract input data from it, and consequently demonstrate untaught behaviours. Intelligent systems require machine intelligence through artificial intelligence algorithms to complete these input data manipulations. However, producing intelligent electronic textiles is a current research challenge. Hypothesising their eventuality and ubiquity, challenges such as remote communication, power generation, data processing, security, and ethics arise. In what remains we focus on the ethical implications and approaches to risk mitigation.

Keywords

Smart textiles, e-textiles, electronic textiles, artificial intelligence, autonomous systems, ethics.

ACM Classification Keywords

• **Human-centered computing**~Ubiquitous computing • **Hardware**~Sensor devices and platforms

Permission to make digital or hard copies of all or part of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. Copyrights for components of this work owned by others than the author(s) must be honored. Abstracting with credit is permitted. To copy otherwise, or republish, to post on servers or to redistribute to lists, requires prior specific permission and/or a fee. Request permissions from Permissions@acm.org.

UbiComp/ISWC '16 Adjunct, September 12 - 16, 2016, Heidelberg, Germany
 Copyright is held by the owner/author(s). Publication rights licensed to ACM.
 ACM 978-1-4503-4462-3/16/09...\$15.00
 DOI: <http://dx.doi.org/10.1145/2968219.2968558>

Introduction

Intelligent systems are being integrated with textile surfaces, changing the perspective of how smart electronic textiles could be. With intelligence – using machine learning, for instance – electronic textiles have the capability of engaging users via feedback and construction of a contextual framework of users' behaviour. However, this position paper argues that transitioning smart textiles to truly intelligent textiles can only be facilitated if they become autonomous systems, and we consider the potential challenges ahead. Likely applications include monitoring and data collection and processing, such as in medical and smart home environments. Through autonomy however, challenges such as remote communication, power generation, security, and ethical considerations arise.

Machine Learning for Intelligent Textile Interactions

Machine learning algorithms have been used to quantify movement of smart textile garments; e.g., to convey different data output compared to its input [1]. Zhang and Harrison used Electrical Impedance Tomography (EIT) to measure the electrical impedance of the inner forearm to decipher hand gestures and finger-thumb pinches in real-time for non-verbal communication to control a smart watch [1]. Computer vision software was programmed to recognise the hand and finger gestures. Support vector machine (SVM) algorithms were used to combine the recognised gestures with the impedance measurements to make decisions on what was communicated. Machine learning algorithms that can build intuitive relationships between data, based on the categorization of this extracted data, can change user's perspective on interaction intelligence. Simple gestures with slight variability could be made distinct

due to SVM classification [1], allowing the EIT system to recognise 378 additional features of the gestures – evidencing the advantage of machine learning to introduce more context and understanding to pre-defined conditions. Arguably, artificial intelligence algorithms used by intelligent textiles can provide understanding about the data it receives. Autonomous behaviour is important in this context as the intelligent textile would showcase independent and trusted decision-making. Autonomous actions based on a deep understanding of the data it handles can improve its user's actions, environment, and relationships with other 'connected' objects if connected to The Cloud.

Other uses of machine learning in wearable systems include the support of assisted living in smart homes. Smart textiles are gaining preference over wearable hardware devices for their compactness, softness, and flexibility [2]. A smart textile made from piezo-resistive material was worn on a finger to detect changes in pressure and strain [2]. Via a smartphone app, the system could recognise flexes of the piezoresistive finger sleeve with high precision to provide an eye-free interaction and control of connected devices. This technology could widen the user-demographic of the assisted living industry, and potentially enable those with physical and mental impairments to control their environment with greater ease.

Personal Interactivity with Intelligent Textiles

For a textile to be intelligent, it must display decision-making attributes, which involves behavioural analysis based on input data and communicate it to its user(s) [3]. Interacting with intelligent textiles can provide helpful behavioural feedback and information about its

environment – which can promote engagement, quantity of collected data, and the increase of contextual understanding. Machine learning techniques have been used in the ‘affective computing’ domain to attain more knowledge on how emotions form and change over time [4]. Such knowledge can reveal physiological and mental states of an individual in both intensity and whether the state is positive or negative [5]. Nardelli used heart rate variability (HRV) as an indicator of emotional arousal and valence to categorise emotions quantitatively. Specifically, musical sounds were used to encourage emotional responses and feature selection algorithms, such as the Leave-In-Leave-Out, were further used to teach the machine intelligence within the system to eventually create an automated classification of emotion recognition. If further interaction with an intelligent smart textile increased its intelligence, it could build its knowledge [6] and begin to extrapolate how existing data would be applied to new contexts [7].

Furthermore, accelerometer and gyroscope sensors can be embedded into intelligent smart textiles to gain data through movement. Accelerometer data analysed by machine learning algorithms such as SVMs, Naïve Bayesian Networks, and other instance-based learning algorithms can also be applied to intelligent smart textile bed mattresses for applications such as acquiring sleep posture patterns [8]. Emotions during sleep can affect quality of sleep and subsequently overall wellbeing [9]. Using intelligent smart textiles in this way, with the possibility to gain more personalised intelligence through radio frequency identification (RFID) tagging to aid assisted living in different locations [10], could help formulate a wellness sensor network [11]. This would create a wireless collective of

shared knowledge within the network, improving conclusions, knowledge, and judgements to increase levels of intelligence and autonomy.

Final Considerations

Deploying digital governance [12] on the intelligent behaviours produced by intelligent smart textiles, requires attention. Necessary security measures to protect those interacting with these textiles are imperative for ethical reasons – to establish what they can reveal and to whom. Moreover, they require reciprocal interactions [13] which are dependent on usability, user-acceptance [14], and perceived trustworthiness [15]. However, as interactivity becomes invisible [16], a ubiquitous data security framework is needed. The extent the user can decide which behaviour they want monitoring needs to be defined. Additionally, how the deductions the machine learning algorithms - which gives the smart textile intelligence - makes about the user will be used by an external party.

In conclusion, intelligent smart textiles will enable soft, flexible, and novel interfaces. Achieving these truly integrated intelligent smart textiles with unobtrusive electronics requires the establishment of reliable manufacturing methods [17]. This paper provides a vision of autonomous intelligent electronic textiles supporting assisted living. Contextual information can be added via location tagging, RFID, and with more interaction patterns can be enhanced by the machine intelligence to create personalised feedback. Though, if intelligent textiles could demonstrate autonomy in the future there are risks and challenges to consider. Further, the efficiency of autonomous intelligent smart textiles requires sufficient user-interaction to give

machine learning algorithms enough data to learn about its environment and user. Additionally, the security implications on what the collected data can be used for and with. Nevertheless, the potential creation of intelligent smart textiles is this – with artificial intelligence capabilities it could make textiles autonomous ‘computational skin’ on a body or object. Potentially, this could turn any interaction with a textile into an intelligent communication of data.

References

- [1] Zhang, Y. & C. Harrison (2015). Tomo: Wearable, Low-Cost Electrical Impedance Tomography for Hand Gesture Recognition. Proc. of the 28th Annual ACM Symposium on User Interface Software & Technology.
- [2] Yoon, S. H. (2016). Wearable textile input device with multimodal sensing for eyes-free mobile interaction during daily activities. *Pervasive and Mobile Computing*, In Press, Available online 29 April 2016.
- [3] Bai, Z. Q. et al (2015). Connexion: Development of interactive soft furnishings with polymeric optical fibre (POF) textiles. *Int.J. Clothing Science and Technology*.
- [4] Matiko, J. W., Wei, Y., Torah, R., Grabham, N., Paul, G., Beeby, S., & Tudor, J. (2015). Wearable EEG headband using printed electrodes and powered by energy harvesting for emotion monitoring in ambient assisted living. *Smart Materials and Structures*, 24(12), 125028.
- [5] Nardelli, M. V. (2015). Recognizing Emotions Induced by Affective Sounds through Heart Rate Variability. *Affective Computing*, 6(4), 385-394.
- [6] Davis, F. (2015). The Textility of Emotion: A Study Relating Computational Textile Textural Expression to Emotion. In Proc. of the 2015 ACM SIGCHI Conference on Creativity and Cognition, 23-32.
- [7] Picard, R. W. (2000). *Affective Computing*, MIT Press.
- [8] Nuksawn, L. N. (2015). Real-time sensor-and camera-based logging of sleep postures. In 2015 Int. Conf. Computer Science and Engineering (ICSEC), 1-6.
- [9] Bianchi, A., Mendez, M.O., Cerruti, S. (2010). Processing of signals recorded through smart devices: sleep-quality assessment. *IEEE Trans. Information Technology in Biomedicine*, 14, 741-747.
- [10] Parada, R., et al. (2015). Using RFID to detect interactions in ambient assisted living environments. *Intelligent Systems*, 30(4), 16-22.
- [11] Ghayvat, H. Liu, J., Mukhopadhyay, S. & Gui, X. (2015). Wellness Sensor Networks: A Proposal and Implementation for Smart Home for Assisted Living. *Sensors*, 15(12), 7341-7348.
- [12] Williamson, B. (2014). Knowing public services: Cross-sector intermediaries and algorithmic governance in public sector reform. *Public Policy & Administration*, 29(4), 292-312.
- [13] Vega-Barbas, M., Pau, I., Ferreira, J., Lebis, E. & Seoane, F. (2015). Utilizing Smart Textiles-Enabled Sensorized Toy and Playful Interactions for Assessment of Psychomotor Development on Children. *Journal of Sensors*, Volume 2015, Article ID 898047.
- [14] Pai, F. Y. & Huang, K.I. (2011). Applying the Technology Acceptance Model to the introduction of healthcare information systems. *Technological Forecasting and Social Change*, 78(4), 650-660.
- [15] Vega-Barbas, M., Pau, I. & Seoane, F. (2014). Confidence: dependencies and their critical role in fostering user acceptance in pervasive applications. *Proc. of Mobihealth'14*, 283-286.
- [16] Poupyrev, I. G. (2016). Project Jacquard: Interactive Digital Textiles at Scale. In Proc. of the 2016 CHI Conference on Human Factors in Computing Systems, 4216-4227.
- [17] Functional Electronics TexTiles (2015), EPSRC Grant EP/M015149/1. <http://www.fett.ecs.soton.ac.uk/>