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Multimodal Data Collection for Social Interaction Analysis In-the-Wild

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Figure 1: Example Snapshots of a mingling event. Taken from the MatchNMingle Dataset [2]

ABSTRACT

The benefits of exploiting multi-modality in the analysis of human-human social behaviour has been demonstrated widely in the community. An important aspect of this problem is the collection of data-sets that provide a rich and realistic representation of how people actually socialize with each other in real life. These subtle coordination patterns are influenced by individual beliefs, goals, and, desires related to what an individual stands to lose or gain in the activities they perform in their every day life. These conditions cannot be easily replicated in a lab setting and require a radical re-thinking of both how and what to collect. This tutorial provides a guide on how to create such multi-modal multi-sensor data sets when holistically considering the entire experimental design and data collection process.

CCS CONCEPTS

• Hardware \rightarrow Sensor applications and deployments; Wireless integrated network sensors; • Information systems \rightarrow Social networks; • Human-centered computing

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 \rightarrow Collaborative and social computing devices; • Applied computing \rightarrow Psychology; • Computing methodologies \rightarrow Camera calibration.

KEYWORDS

ConfLab, Social Behaviour Analysis, Wearable Sensors, Multimodal Synchronization

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1 MOTIVATION

Developing automated methods to analyze human social behavior in crowded face to face settings is an important multimedia concern [1, 2, 4]. With the rising importance of mobile and ubiquitous comping, multiple recording modalities are no longer tied to a fixed location in a lab setting. They are distributed in and amongst people as they move around in the world.

Harnessing the ubiquity of these sensing modalities is still an open question. Typically mobile computing applications rely on relatively low resolution data in order to accommodate a reasonable battery life for its users since mobile phones are also used as personal devices. Given the reduction in energy consumption and size of modern day electronics, sensing

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devices are also being embedded into everyday objects such as smart ID badges. Such devices allow for a 'grab and go' paradigm to the ubiquitous sensing idea where consent to be recorded does not require a lengthy process of installing a mobile app or having to compete for processing power with other mobile applications. Such smart ID badges allow a tradeoff between battery life and higher sample rates in on-board sensors, such as those measuring body movement.

Exploiting accelerometer readings of body movement have shown promise for detecting social actions such as speaking activity or for detecting social involvement, but the solutions to such problems are still in their infancy [5, 6]. Multimodal attempts have also shown promise [1, 3] At this stage, video and audio data recorded in conjunction with wearable sensor data is vital for us to understand the relationship between automated social behaviour analysis as a multimodal and as a single modality problem.

Despite interest community interest in this topic, research works still remain relatively few. This could be strongly related to the difficulty of collecting relevant and sufficiently large data sets. Practical advice is needed to help researchers in the Multimedia community to design the data collection process appropriately. It calls upon an interdisciplinary approach to the data collection and experimentation process which is often not part of the standard educational curriculum of many multimedia researchers.

2 COURSE DESCRIPTION

This tutorial is tied closely to a social experiment carried out during ACM Multimedia 2019 called *ConfLab*. ConfLab aims to create a community level data gathering event. It involves recording conference attendees while they socialize as a means of helping attendees and organizers to network. To our knowledge, ConfLab is the first of its kind to turn the data collection process as a tool for community level introspection. The aim of ConfLab is to also make the data collected to be shared under appropriate levels of privacy restriction with the wider research community as a stimulus for grand challenge innovation.

The tutorial also acts as a debriefing and a moment for community reflection where participants or non-participants of ConfLab can share their thoughts on the initiative with others. We particularly encourage both newcomers and experienced members of the community to join. All perspectives of the conference from the attendees are vital for enabling a balanced and diverse learning experience.

This half day tutorial covers many of the practical considerations of collecting data which are often not documented in research papers, being not considered of academic interest for research on the automated human behaviour analysis in the wild. However, the issues highlighted in this tutorial are crucial to consider when collecting data in semi-public spaces. The tutorial is partially based on the book chapter by Hung *et al.*[7] providing more detailed practical advice at all levels of the collection process. In addition, themes more specific the data collection process of ConfLab will also be discussed. The structure of the tutorial is divided into lectures and guided group discussion. The group discussions will primarily be focused on the debriefing of the ConfLab experiment and discussions about data sharing, how to incentivize participation, and how to include a participatory design approach into community data collection through thinking about ethical AI practices. Concretely, the following themes will be covered:

- Theory
 - Defining In-The-Wild vs Ecological Validity
 - Thin slice approach to Behavior Analysis
 - The inductive vs deductive approach when collecting data
- Practical issues (lectures)
 - Wearable sensor design
 - Multicamera video
 - Audio and privacy concerns
 - Multimodal data synchronization
 - Pilot Tests
- Ethical issues (lectures)
 - Applying for ethical approval
 - Informed consent
 - Data sharing (with particular focus on recent European Union laws on the General Data Protection Regulations (GDPR) introduced in 2018.
- Human concerns (group discussion)
 - Debriefing on ConfLab
 - How to incentivize participation: identifying stakeholders and their needs
 - Ethical data sharing practice

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