



Group Activity Recognition using Mobile Devices

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Human / Single-user (HAR) Track activities of a single user over period of time Multi-user (MUAR) Tracking activities of several users in parallel

Solo, cooperative, conflicting activities

Group (GAR)

- Group is an abstract entity: organism
- Emergent behavior based on activities and interactions of individuals: flocking
- GAR is estimation of group behavior classes:
- Analogous to HAR with multiple sensing modalities/locations, but interactions complex



Thesis



 Devices of group members can collaborate to estimate group behavior being generated
Recognition in mobile sensing network solves infrastructure issues





Research Questions



Technology for

Pervasive Computing

Problems

- Distributed execution reducing communication/processing
- Observing identity not just what...but from whom?
- Dynamic membership people come and go







Gordon et al.: Recognizing Group Activities using Wearable Sensors

Technology for **Pervasive Computing**

Distributed Classification – Probabilistic Appr.



- Distributed probabilistic inference (DPI) using belief propagation (BP)
 - Evaluated for distributed sensor calibration
- Usefulness for GAR will be evaluated
 - Human factor: human interaction more complex than sensors
- How DPI-BP works
 - 1) Estimate group activity based on local sensors
 - 2) Exchange estimation with neighbors
 - 3) Re-estimate using local sensors + neighbor estimations
 - 4) repeat from 2
- Converges to "correct" answer
- single-user, multiple sensor location experiments: in progress
- Multiple-user, single sensor location (iPhone): planned



$$\Pr \{\mathbf{X}, \mathbf{M}\} = \underbrace{\left[\frac{1}{Z}\prod_{\mathbf{C}\in\mathcal{C}}\psi_{\mathbf{C}}(\mathbf{C})\right]}_{\text{forterized prime Dr}\left(\mathbf{X}\right)} \prod_{k=1}^{K} \underbrace{\Pr \{M_k \mid \mathbf{B}_k\}}_{\text{measurement model}},$$

factorized prior $\Pr \left\{ \mathbf{X} \right\}$

Paskin & Guestrin.: Robust Probabilistic Inference in Distributed Systems



- Problem: people come and go, each individual affects group behavior differently
 - Experience (prior) learned for one individual may not fit the next one
- Solution, factorize prior using social role of individual
 - Role: characterization of behaving similarly w.r.t. to the group activity
 - Assumption: priors transferrable within role
 - Requires expert knowledge of roles a priori
 - People may change roles (out of scope)

Evaluation:

- Simulate replacement of subjects in a group data set
- Exchanging subjects of same role vs. random
- Crunch zone!





- Role can be seen as user's bias to a group activity
- Extracting a user's role
 - If the group activity is known: infer role
 - Over time, estimations of social role of members improves, converging.
- Using role to improve GAR
 - If role is known: infer activity
 - Knowing a member's bias improves inference of correct group activity

Evaluation:

- Role recognition using human flocking behavior data set
- Roles: leader and followers in a flock
- Planned time line: Sep. through Nov. 2012 at IfE / WCL (Dr. Roggen) ETH





- GAR presents a novel problem
 - Similar to HAR: but complex interactions between members
 - Similar to MUAR: distributed sensing but total is more than sum of individuals
- Complexity of human interactions and dynamic group membership pose issues
- DPI+BP is a promising distributed approach
- Roles can solve dynamic group membership issues
- Expected result: DPI+BP with prior factorized over roles provides generalizable solution for GAR

