

Concept Drift: a Challenge for Intent Inferral using Biosignals



Approach: Disagreement-based Semi-supervision

- We aim to train on a small dataset labeled by the experimenter, then adapt to concept drift using unlabeled data.
- Our algorithm uses an ensemble of classifiers. Each classifier uses a randomly sampled subset of the sensing modalities.
- When concept drift occurs, some modalities remain robust. We can use the more confident classifiers to improve less confident classifiers.



Adaptive Semi-Supervised Intent Inferral to Control a Powered Hand Orthosis for Stroke

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(a) Task Setup

Experiments and Results

(b) Open and Reach

(c) Close and Grasp

(d) Hand Over

Hardware and Sensors Reference IMU Motor position encoder EMG armband IMU sensor Pressure sensor • Our orthosis is an exotendon device

CRA 2022

IEEE International Conference

on Robotics and Automation

- consisting of a forearm splint and fingertip components.
- The orthosis is equipped with multiple sensing modalities: Forearm EMG, Motor Position, Finger Joint Angles, Fingertip Pressure

(e) Open and Place

- We evaluate our method both offline and online.
- Despite being trained only on one condition, our method outperforms the the supervised learning baseline that is trained on all conditions.
- In the online functional task, two subjects successfully completed multiple instances of a pick-and-handover task.