

Recognition of Continuous Mouse Gesture Sequences

I. PROBLEM STATEMENT

Gestures, ranging from simple mouse movement to complex full-body motion, are a common form of input in human-centric user interfaces [1]. However, a significant barrier to widespread use of gesture input is the problem of accurate gesture recognition, both in isolation and in sequence. Gesture recognition frequently involves solving two problems: segmentation of a gesture sequence and recognition of individual gestures.

In our work, we focus on the problem of segmenting sequences of user-defined symbols created via mouse gestures. We strive to be part of a larger endeavor of developing a medium of communication for deaf people who are unable to interact through sign language due to partial paralysis or other similar conditions. Although a mouse may not be the most viable medium for such affected people to use to communicate with others, our work may be suitably enhanced/extended for more accessible mediums of interaction.

II. RELATED WORK

Yang et al [2] present work on recognition of individual gestures in continuous gesture sequences, but unlike our work, they train Hidden Markov Models (HMMs) on continuous gestures; we view this as a deficiency due to the *a priori* definition of gesture sequences. The work most closely related to our problem and approach is focused on recognition of Chinese handwriting [3]. Hong et al use an iterative segmentation technique that uses whitespace separation to split character sequences into individual characters. Their approach is similar to ours, but their approach is intended to be suitable for both off-line and on-line recognition. As a result, their approach does not take advantage of the temporal data that we employ. To our knowledge, there is no directly comparable system in the literature. However, we will compare our results to those presented by Hong et al [3] due to the similarities between our problem and approach.

III. PROPOSED IMPLEMENTATION

Our problem is segmentation of user-defined gestures. As a result, we cannot leverage grammar or other linguistic features, as was done successfully by Starner et al [4]. Additionally, we cannot assume the existence of any markers that separate gestures, as is typically found in Chinese writing [3]. As a result, we employ an approach based on an efficient "search" over the space of segmented gesture sequences, aiming to find the most accurate gesture segmentation.

We will build upon the mouse gesture recognition system developed by Tanguay [5], which performs on-line recognition of individual mouse gestures via HMMs. We will enhance his system by adding a segmentation routine that accurately splits a continuous gesture sequence.

In our approach, we first split the multi-gesture data stream temporally into N sub-streams (each corresponding

to an individual gesture) of identical duration equal to the mean time taken to draw an individual gesture (calculated from the training set). The duration of each sub-stream can be conceptualized as a "window" overlaid on the complete data stream; N sub-streams correspond to N windows. After configuring the initial "window set", we calculate a "window set score", which is the mean of all the individual window scores; each window score, in turn, is equal to the maximum of the probabilities of a match between the window data and each trained HMM. We subsequently iteratively increase the window set score by independently increasing or decreasing the duration of each window by a delta based on the time variance of gestures in the training set. If the window score increased between iterations k and $k-1$, we repeat the previous action; otherwise, we try the other action. Our routine terminates when the delta change reaches zero.

IV. PROPOSED EVALUATION

We propose to evaluate our routine by measuring the success rate for segmenting a gesture sequence into the correct gestures. We define the success rate as the percentage of correctly identified individual gestures. As noted above, our problem and segmentation routine is comparable with the work completed by Hong et al [3]. As a result, we will compare our success rates with their published results.

To avoid author bias during evaluation, we will conduct a micro-study involving one user. The user will arbitrarily create gesture sequences of lengths 1, 2, and 3. Gestures with length 1 serves to validate the system accuracy for individual gestures. Gestures of lengths 2 and 3 serves to quantify the success rate of our segmentation routine.

V. REFERENCES

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