

# A Novel Sequential Monte Carlo Framework For Predicting Ambiguous Emotion State

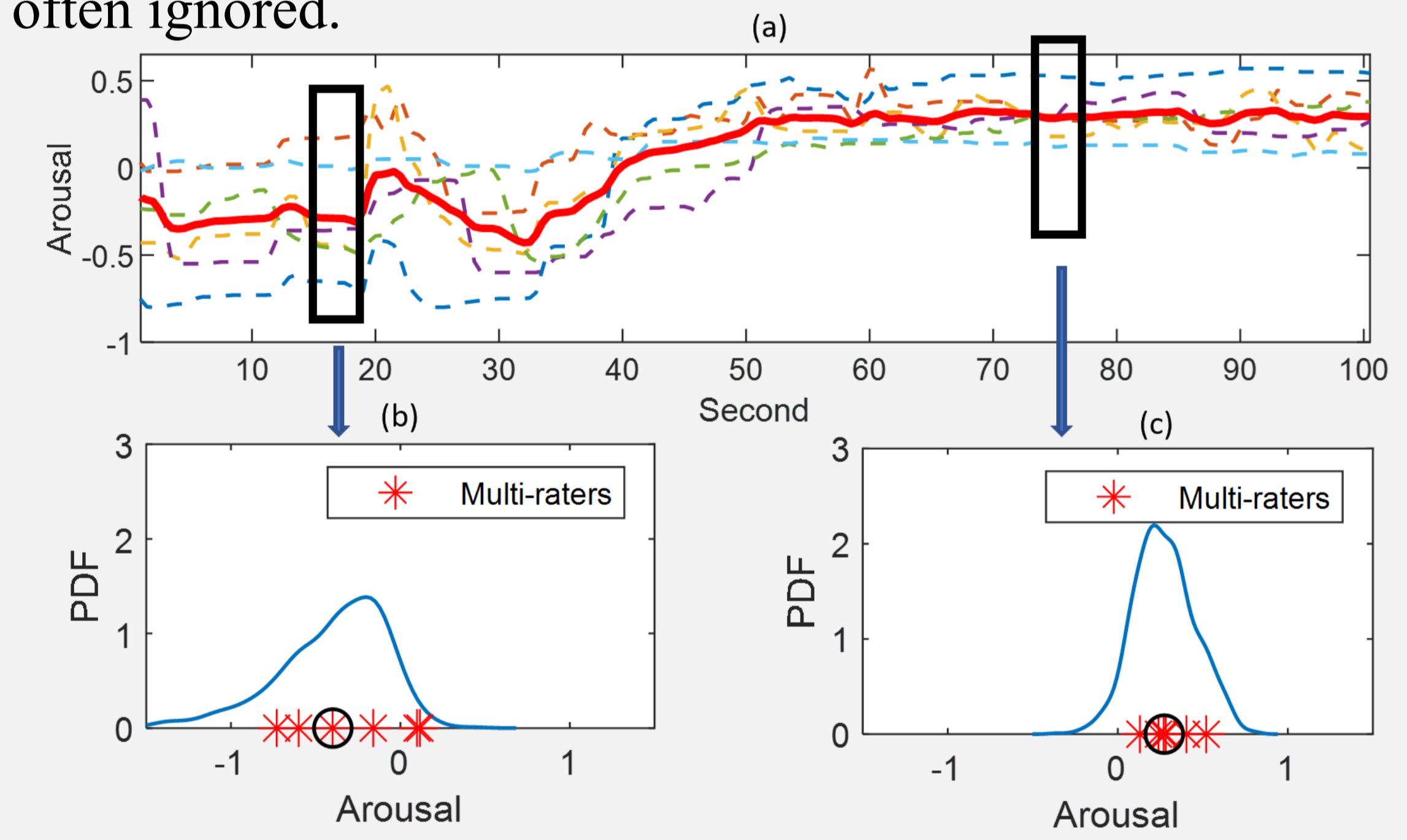
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## 1. Introduction

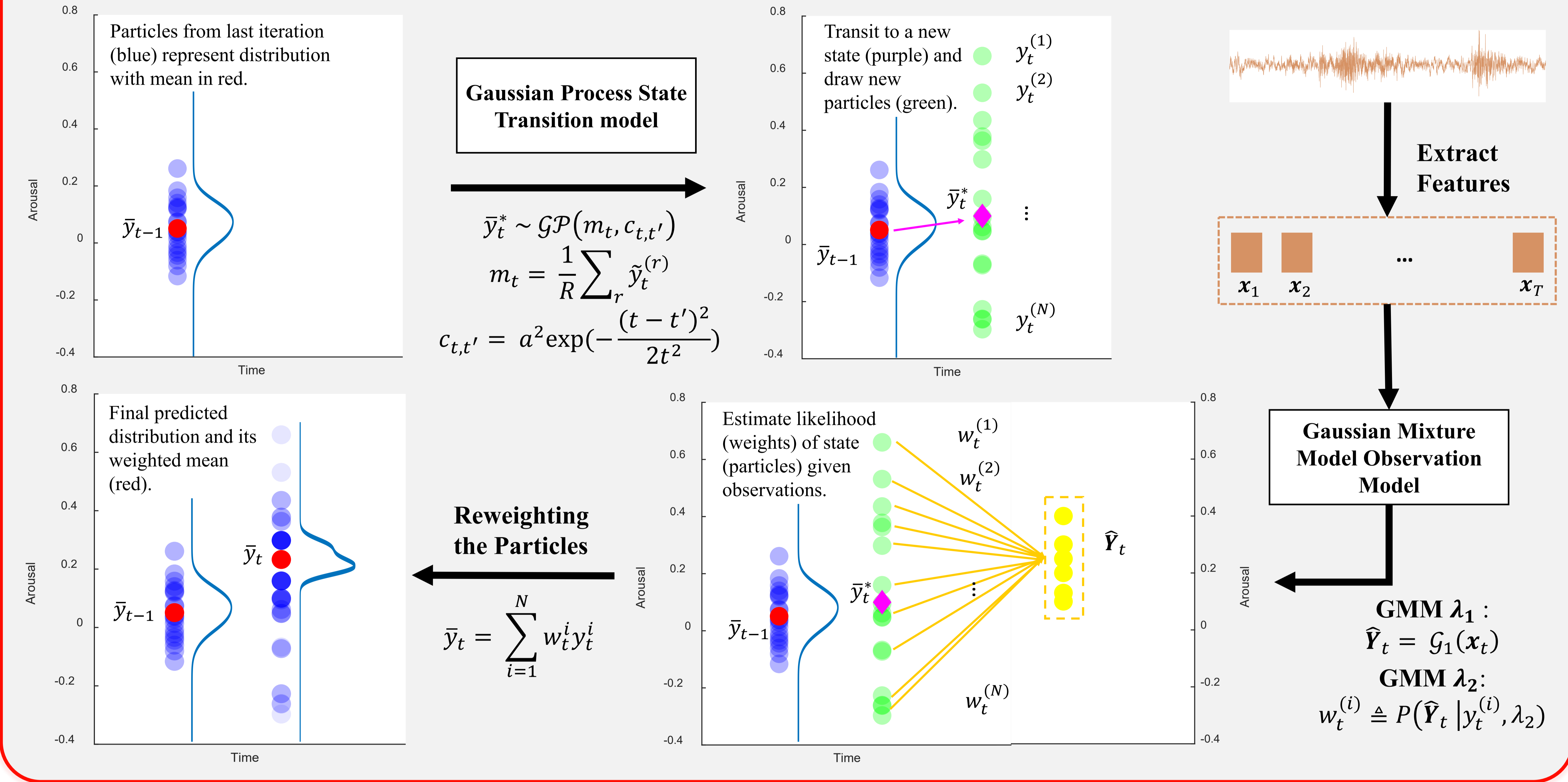
Emotion labels are typically obtained from multiple annotators. However, most automatic emotion recognition systems **ignore disagreement between annotators which reflects the ambiguity and subtlety of emotions**. In this paper, we **propose a novel Sequential Monte Carlo framework** that models the perceived emotion as time-varying distributions that allows for ambiguity to be incorporated.

### Limitations of Current Systems

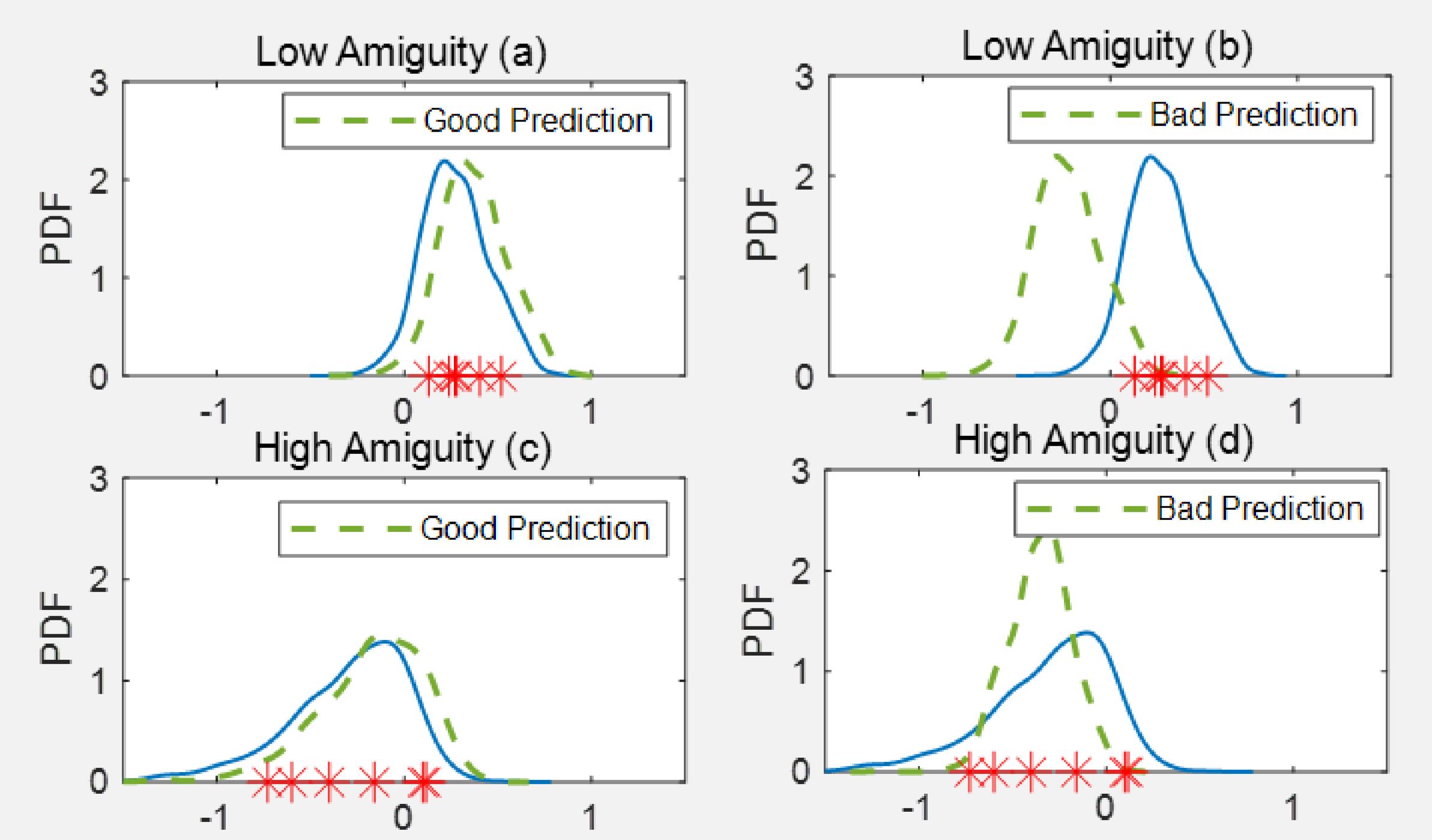
- Typically, only the **mean** rating (solid red line) is considered and inter-rater differences are ignored.
- When distributions are considered,
  - they are often assumed to be Gaussian.
  - temporal dependencies** between consecutive predictions are often ignored.



## 2. Proposed Sequential Monte Carlo Framework

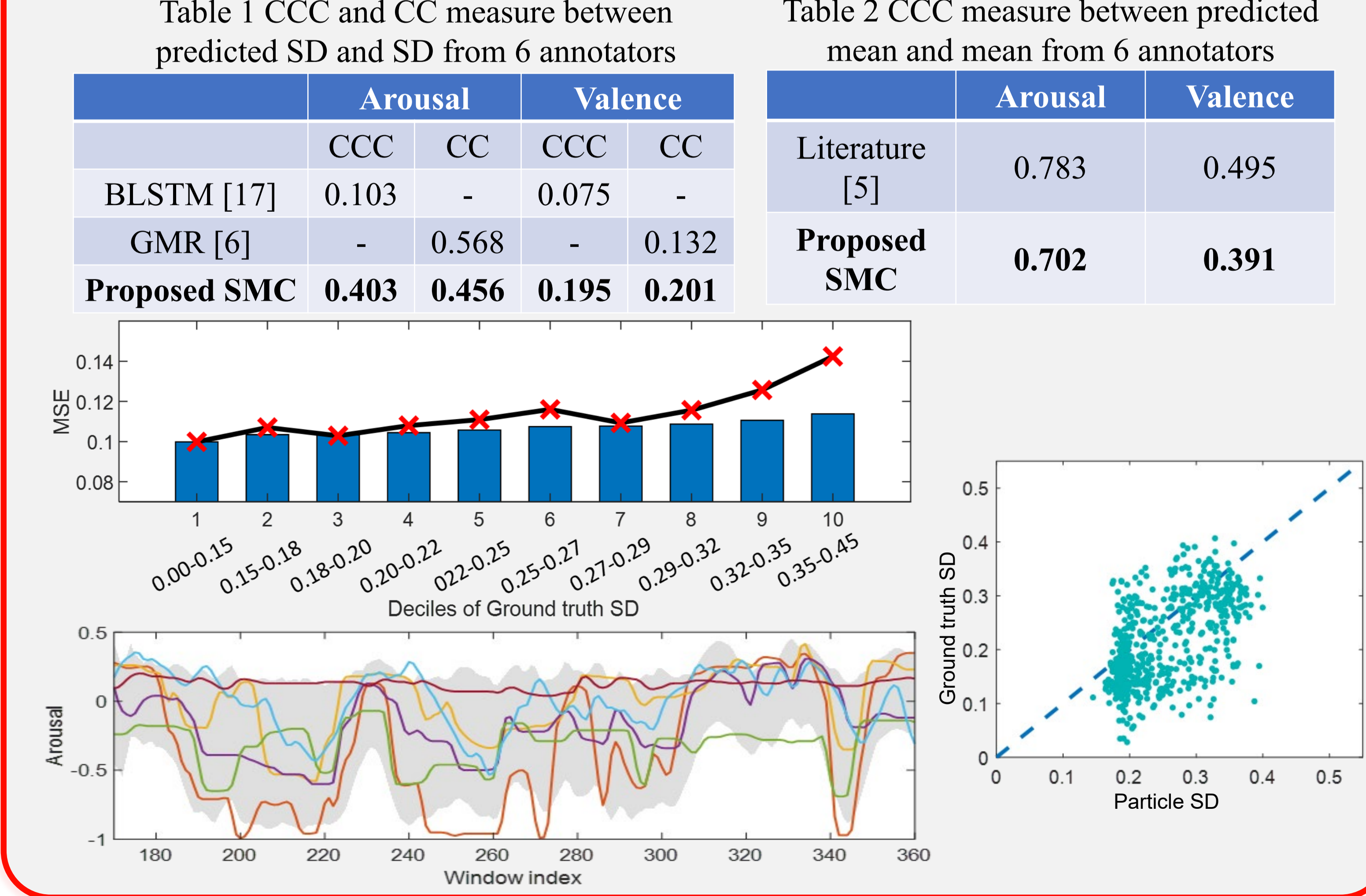


## 3. Proposed Measures



- Concordance correlation coefficient (CCC) between prediction standard deviation the standard deviation of the emotion ratings;
- Mean squared error (MSE) between prediction mean and ground truth mean label for different frames partitioned into deciles based on the standard deviation of the labels.

## 4. Results



## 5. Experimental Settings

- Corpus: the RECOLA dataset; 9 training & 9 development utterances.
- Arousal & valence labels; 6 annotators.
- 40ms sampling rate; 1 second window (50% overlap).
- Delay compensation: 4 seconds for arousal and 2 seconds for valence.
- 8 - mixture GMM for  $\lambda_1$ , 4 - mixture GMM for  $\lambda_2$ .
- 1000 particles.

## 6. Conclusion

- Inter-rater differences in emotion annotations reflecting ambiguity in the state can be represented as a series of non-parametric distributions.
- The proposed framework is able to track the level of ambiguity in the labels over time and predict the emotion state accurately within regions of low ambiguity.