

# Evaluating Tooth Brushing Performance With Smartphone Sound Data

JOSEPH KORPELA<sup>1</sup> • RYOSUKE MIYAJI<sup>1</sup> • TAKUYA MAEKAWA<sup>1</sup>  
KAZUNORI NOZAKI<sup>2</sup> • HIROO TAMAGAWA<sup>2</sup>

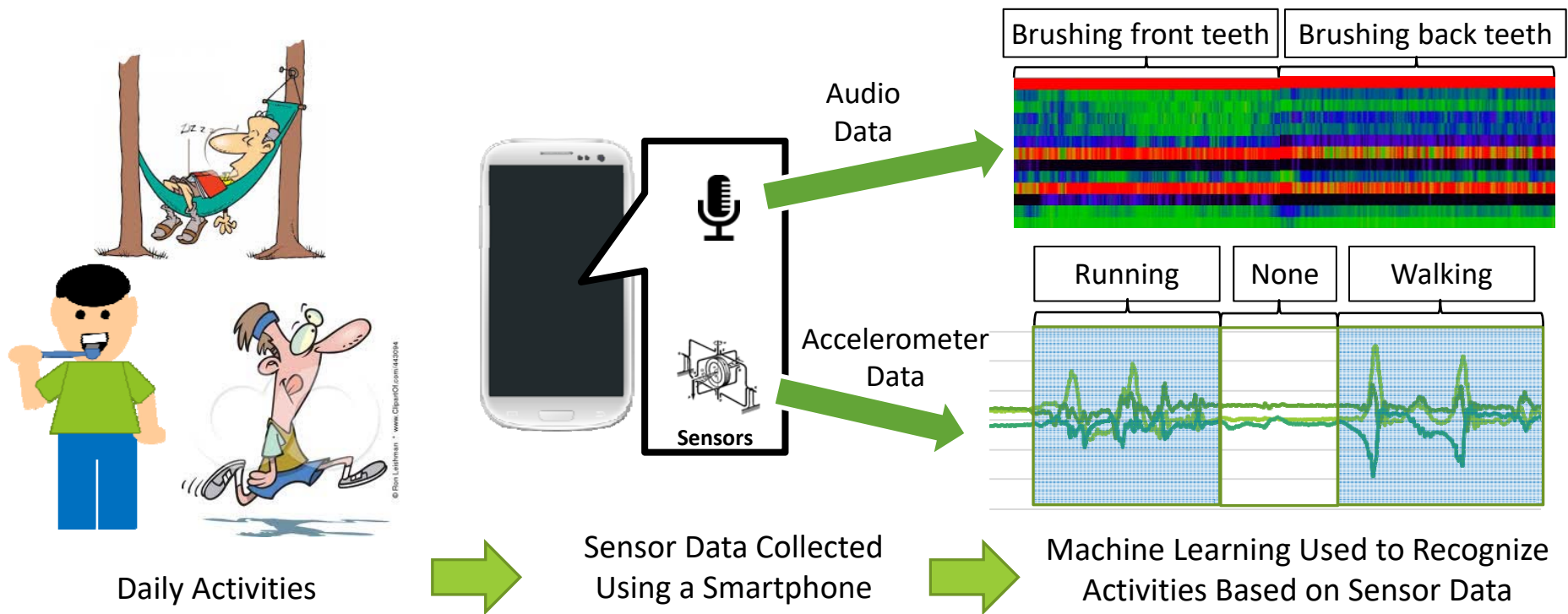
---

<sup>1</sup>OSAKA UNIVERSITY, GRADUATE SCHOOL OF INFORMATION SCIENCE AND TECHNOLOGY

<sup>2</sup>OSAKA UNIVERSITY DENTAL HOSPITAL



# Activity Recognition



# Activity Recognition in Health Care

---

Tracking sleep quality/quantity



Tracking medication intake



Tracking exercise



Tracking food intake



# Dental Health

---

Teeth are important to our health

- Need to last a lifetime
- Tooth loss leads to loss of appetite and decreased nutrition

Brushing is important for our teeth

- Proper brushing improves dental health
- Improper brushing can damage teeth and gums



Yet, most people don't brush well enough



# Activity Recognition for Dental Health

---

Significant improvement in brushing habits when provided feedback via activity recognition techniques<sup>1</sup>

Previous methods have required specialized equipment

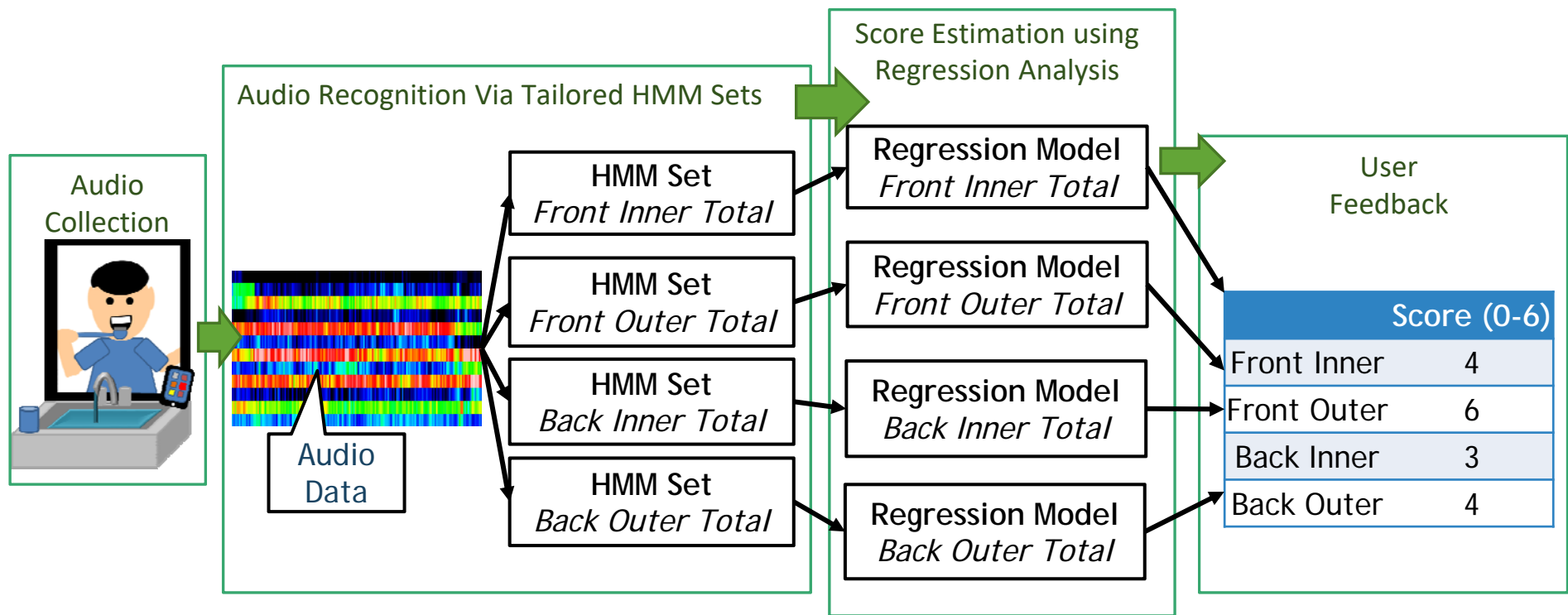
- LED extension for toothbrush<sup>1</sup>
- Accelerometer extension toothbrush<sup>2</sup>



Our method uses only *audio* data:  
Allows users to evaluate brushing using an off-the-shelf smartphone

1. Chang, Y.-C., Lo, J.-L., Huang, C.-J., Hsu, N.-Y., Chu, H.-H., Wang, H.-Y., Chi, P.-Y., and Hsieh, Y.-L. Playful toothbrush: ubicomp technology for teaching tooth brushing to kindergarten children. In *CHI 2008* (2008), 363–372.
2. Graetz, C., Bielfeldt, J., Wolff, L., Springer, C., Fawzy El-Sayed, K. M., Salzer, S., Badri-Hoher, S., and Dorfer, C. E. Toothbrushing education via a smart software visualization system. *Journal of Periodontology* 84, 2 (2013), 186–195.

# Proposed Method



# Evaluation Scores: Plaque Tests

---

## Evaluation Scores

- Regression models need scores to use as training data

## Plaque Test

Typical method of evaluating tooth brushing effectiveness

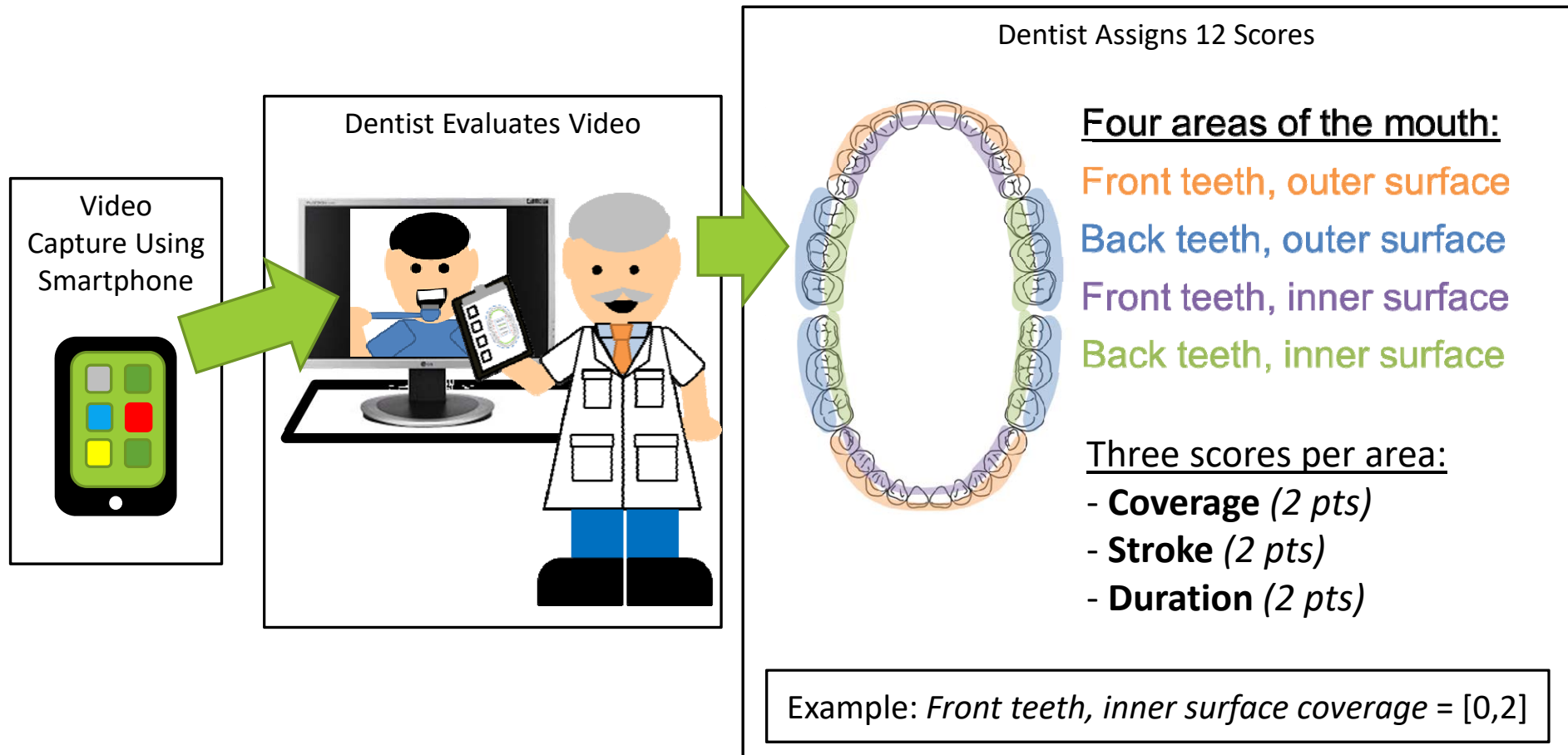
1. Apply plaque indicator liquid to teeth
2. Liquid makes plaque easily visible
3. Dentist evaluates based on plaque left remaining

## Issues with using plaque test

- Influenced by all tooth brushing performed over last few days
- Influenced by foods/drinks recently consumed
- Costly to gather a large number of scores



# Evaluation Scores: Video-based





# Video-based Scores vs. Plaque Test Scores

---

14 Subjects

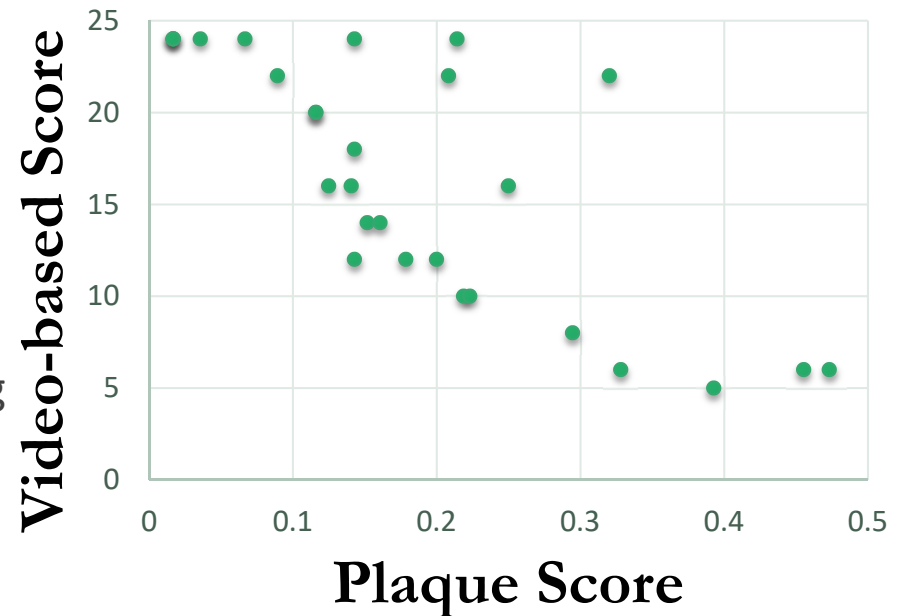
Day 1:

- Brushed teeth with video recorded
- Received plaque test

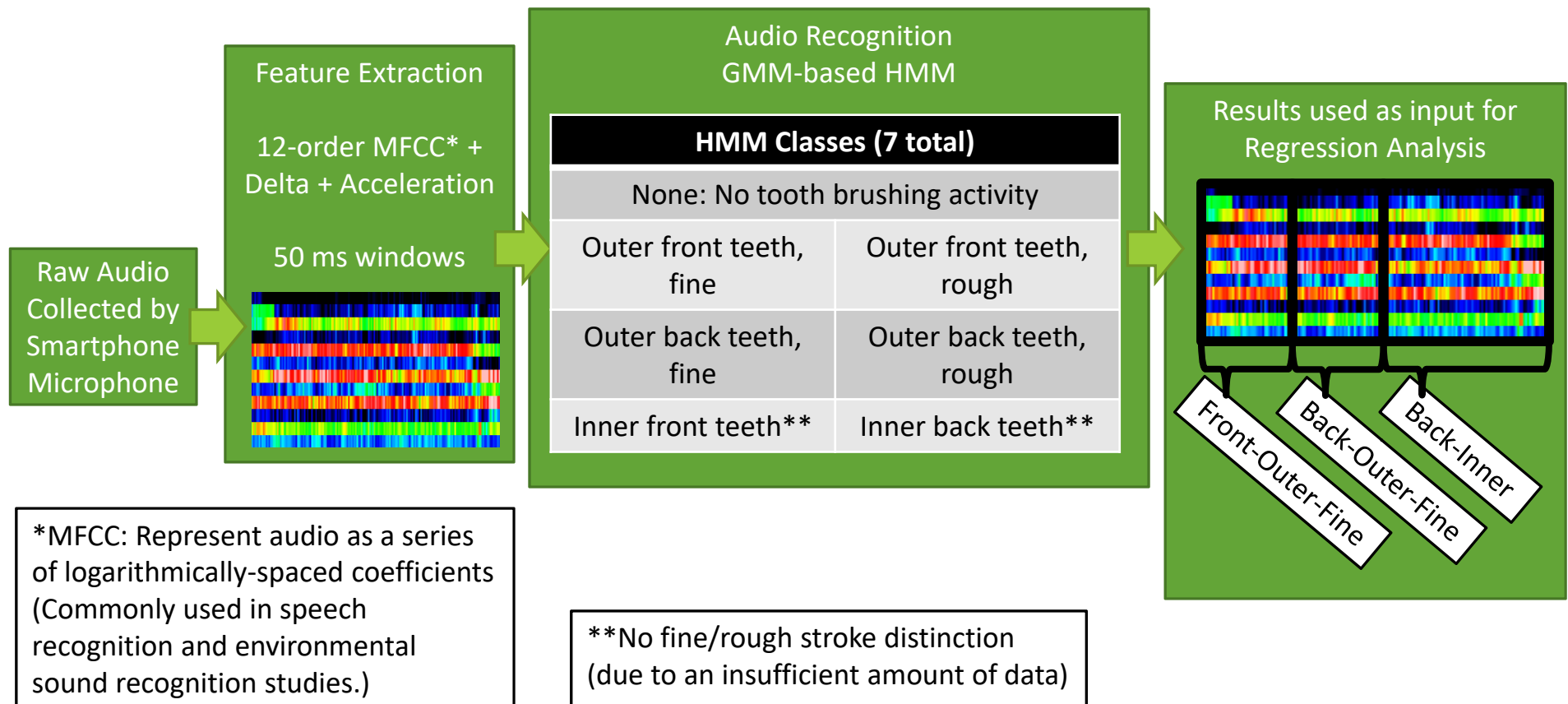
Day 2:

- Received instruction on proper brushing technique
- Brushed teeth with video recorded
- Received plaque test

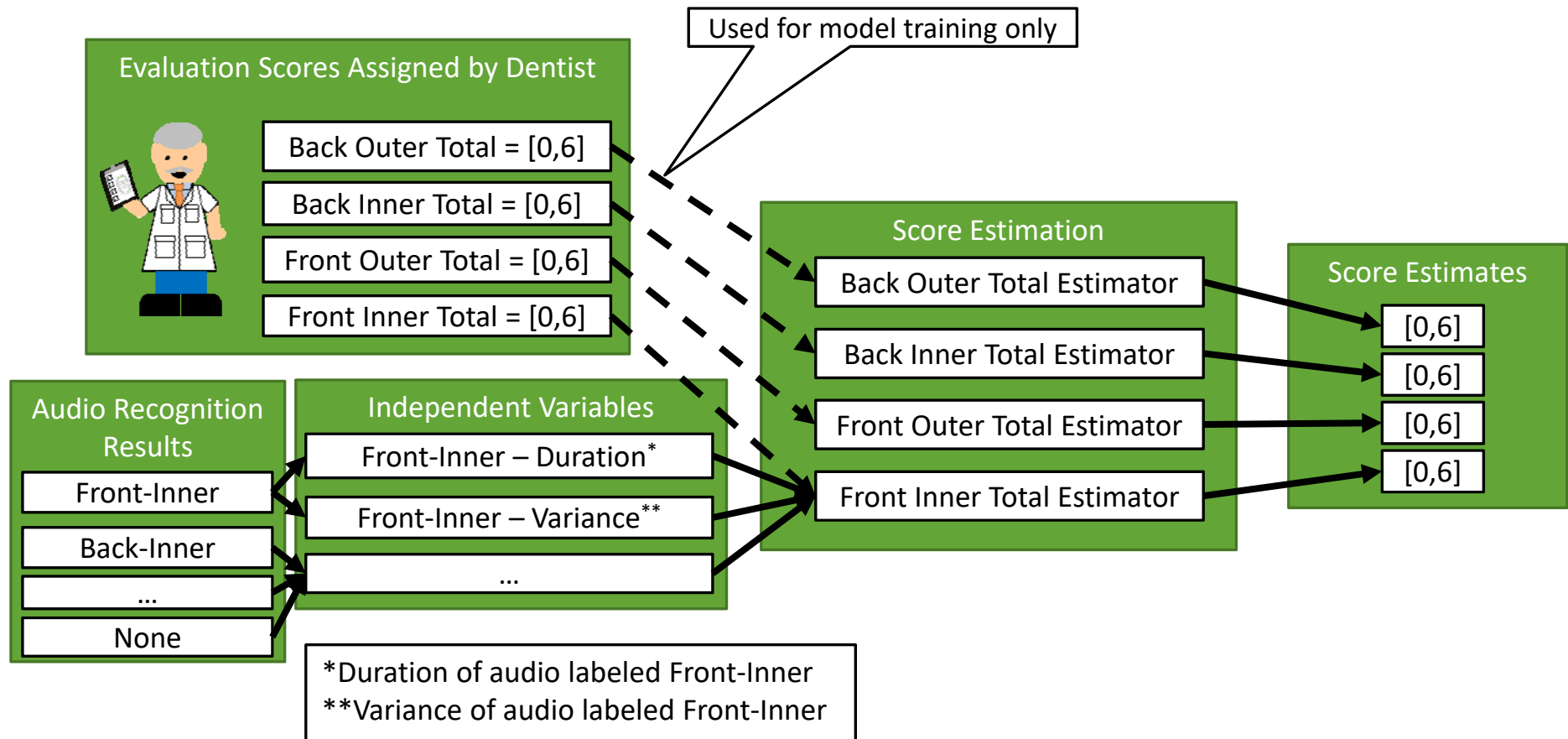
Video data was then used to generate scores for each session



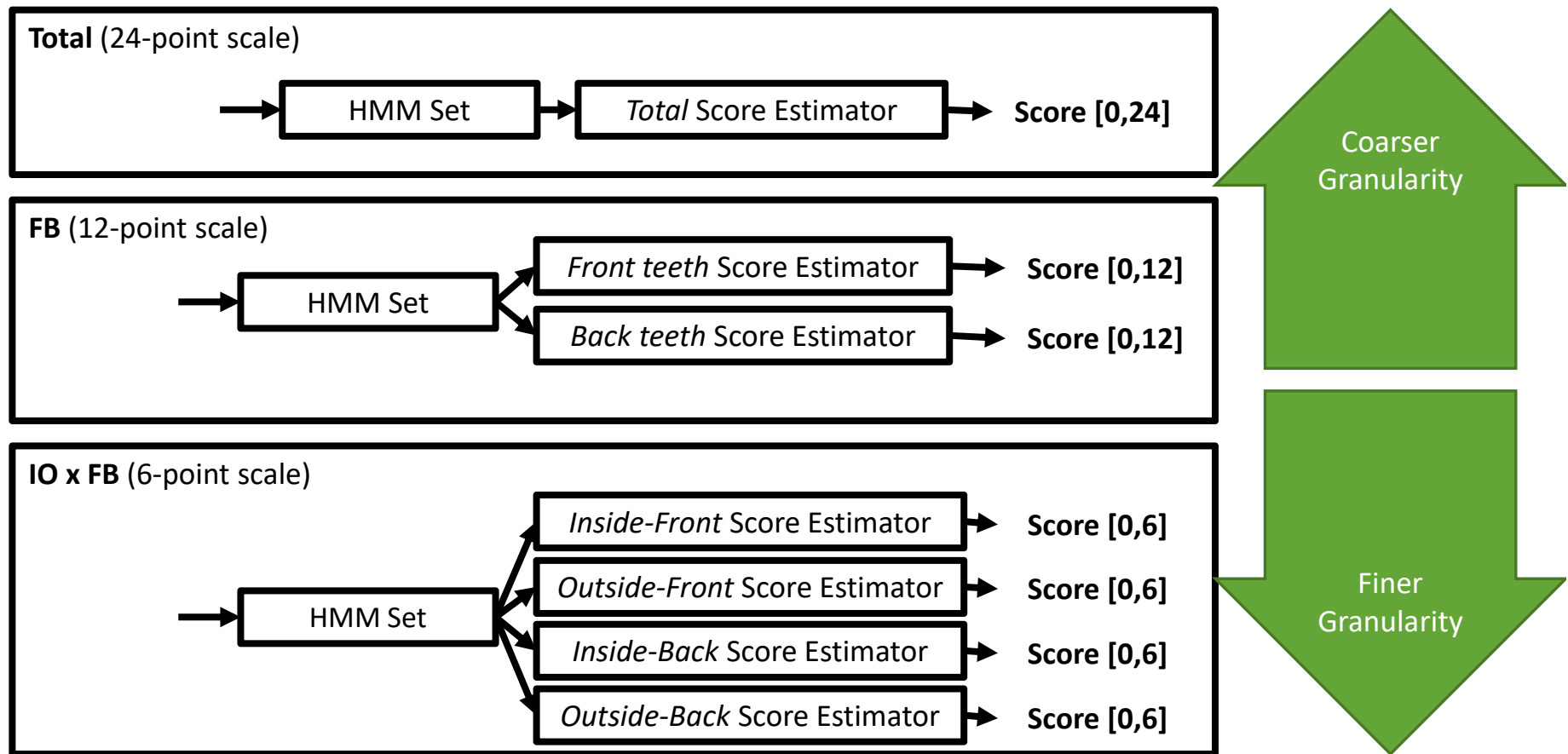
# Audio Recognition



# Score Estimation



# Score Architectures

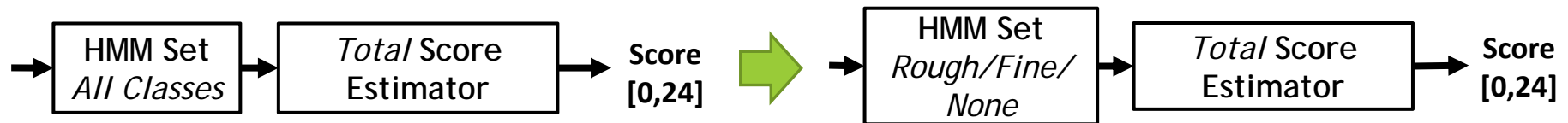


# Improving HMM Performance

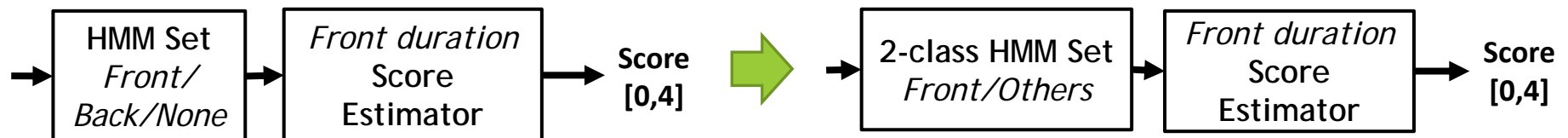
Audio recognition performance is better at coarser granularities

(Accuracy when using all classes: 45.1% → when using only 3-classes: 68.4%)

1. HMM granularity required depends on the score granularity



2. Individual scores require different sets of HMMs

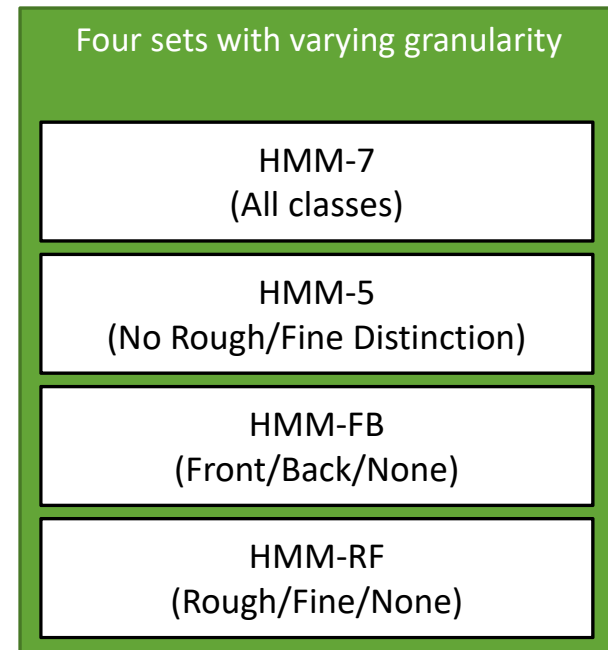
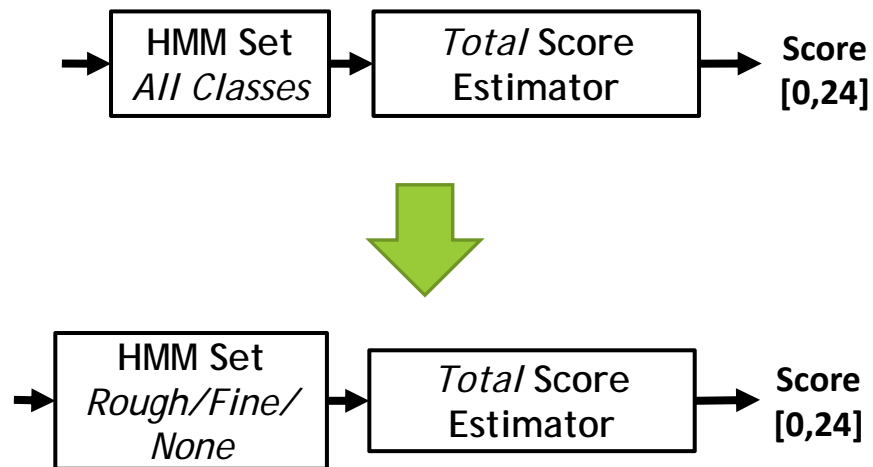


Improving Performance:

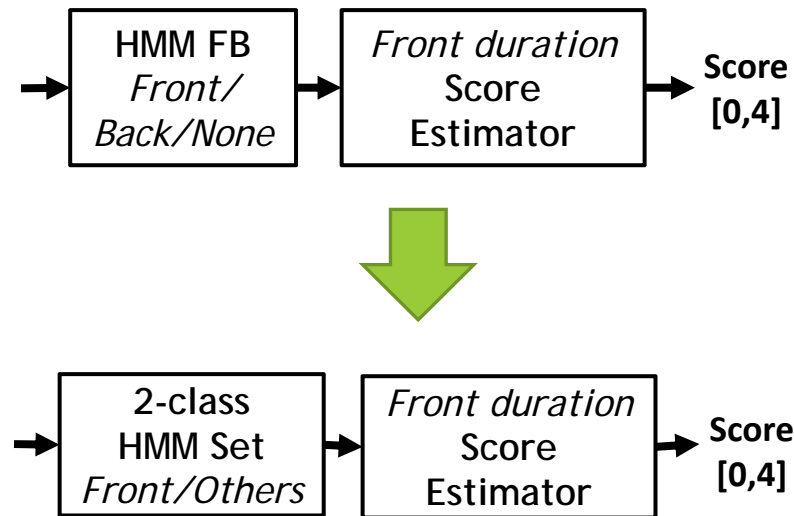
1. Create HMM sets with varying granularity
2. Create HMM sets that are tailored to each score

# Varying HMM Granularity

---



# Tailoring HMM Sets to Regression Scores



Choosing the Most Useful HMM Classes

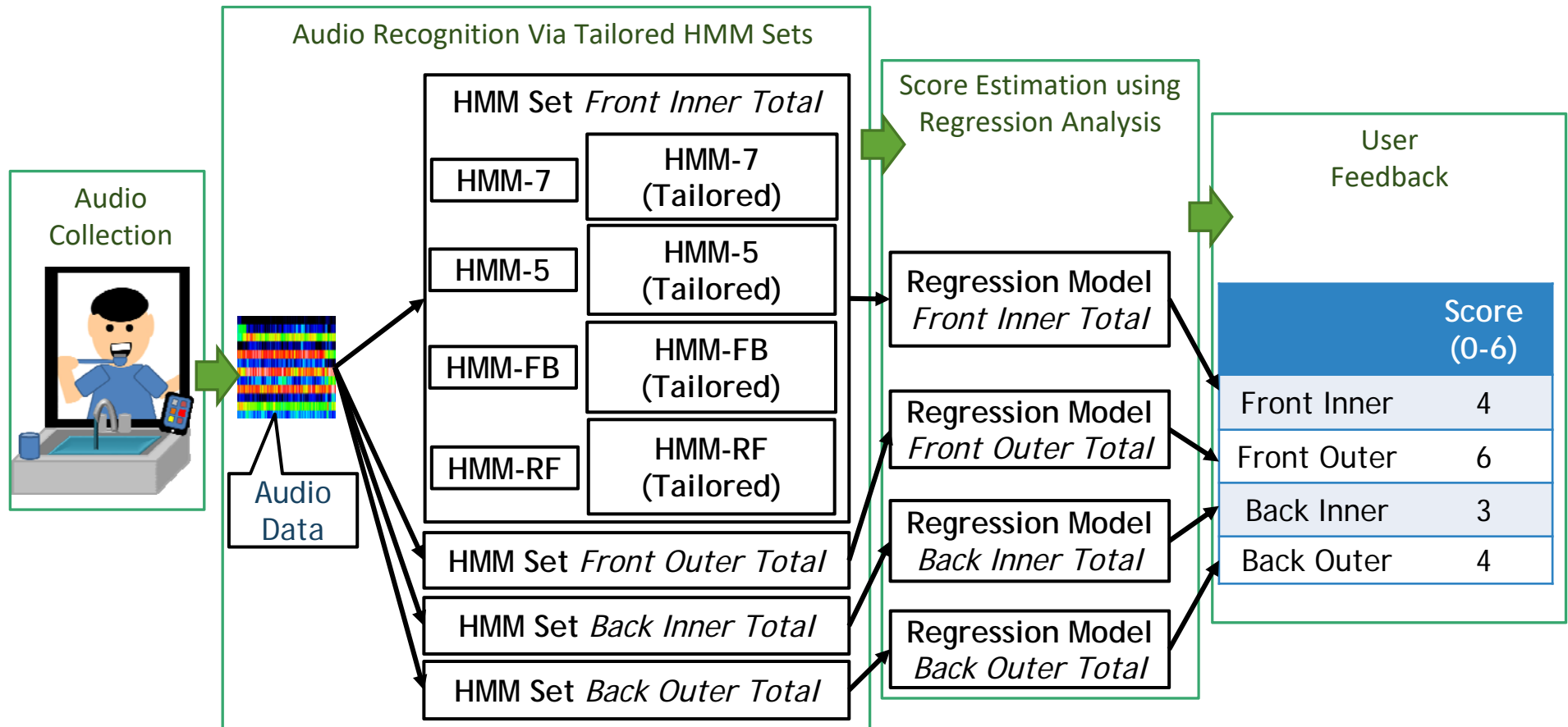
The process starts with an 'Initial full HMM set' box, which leads to 'Generate independent variables', then to 'RReliefF: Calculate a weight for each variable', and finally to 'Choose useful classes based on weights'.

| Variable       | RReliefF Weight |
|----------------|-----------------|
| Front-Duration | 0.4             |
| Front-Variance | 0.25            |
| Back-Duration  | 0.15            |
| Back-Variance  | 0.1             |
| None-Duration  | 0.05            |
| None-Variance  | 0.05            |

| Class | Total Weight |
|-------|--------------|
| Front | 0.65         |
| Back  | 0.25         |
| None  | 0.1          |

# Proposed Method





# Evaluation Methodology

---

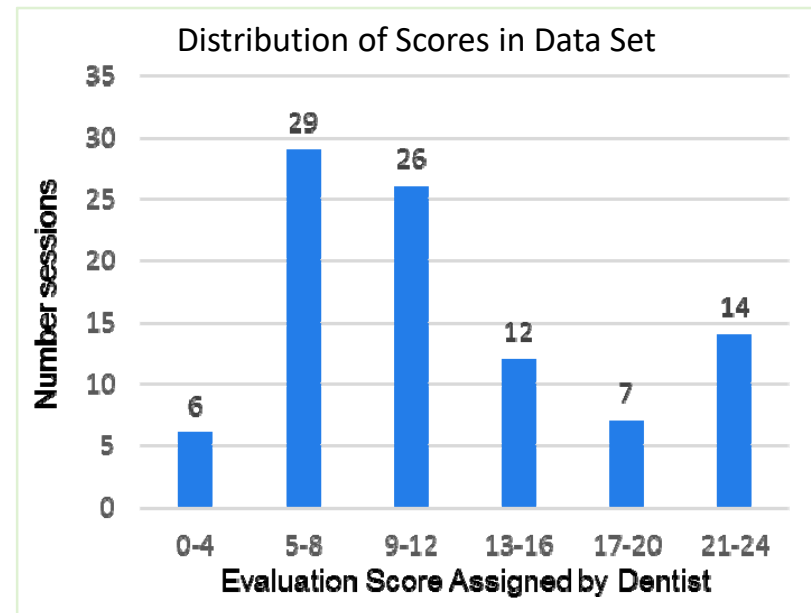
## Data Set

- 94 sessions total
- 14 participants
- Average length of each session: 94 seconds

## Environment

- Collected either in our lab or in the participant's own home
- Users allowed to use own toothbrush or one provided by us

Evaluated using *leave-one-user-out* cross validation



# Score Estimation Evaluation: Methods

---

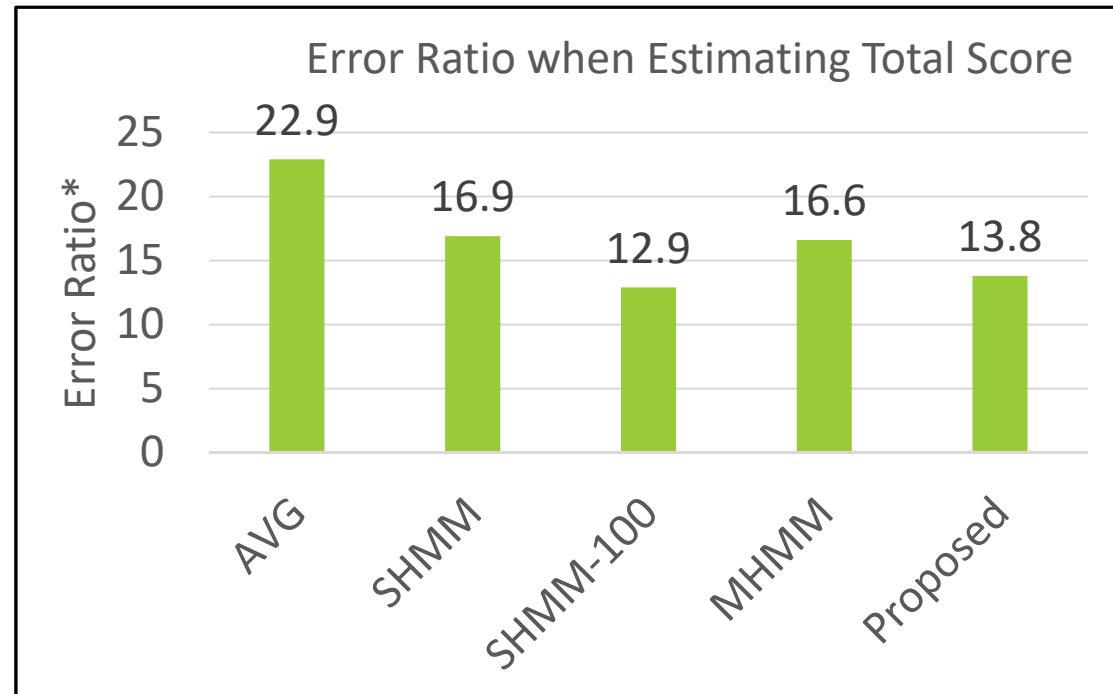
- 1. Avg:** Each user's scores are estimated using the average scores for other users.
- 2. SHMM:** The same HMM set (*HMM set 7*) is used to generate independent variables for estimating all scores.
- 3. SHMM-100:** A variation of the *SHMM* method in which we built the regression models using corrected labels, i.e., this method assumed 100% recognition accuracy for *HMM set 7*.
- 4. MHMM:** Four basic HMM sets: *HMM set 7*, *HMM set 5*, *HMM set FB*, and *HMM set RF*, are used to generate independent variables for estimating the scores.
- 5. Proposed:** The proposed method, in which we prepared a tailored group of HMM sets for each of the scores.

# Score Estimation Evaluation: Total Architecture

---

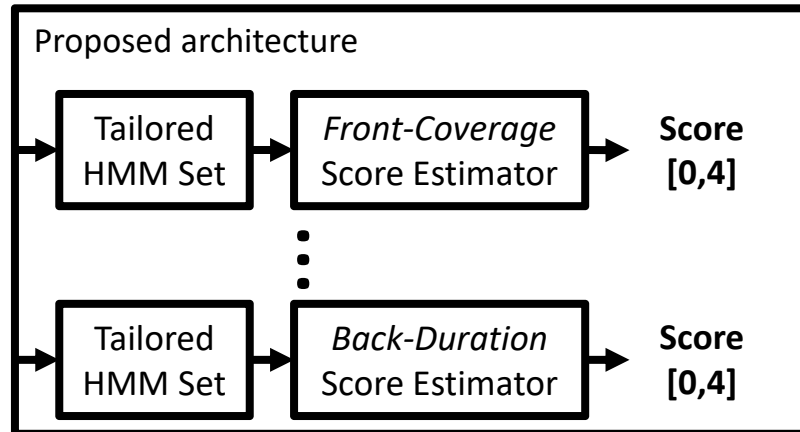


Estimated a single score (24-point scale) that represents the total score for all tooth brushing activity in the session.

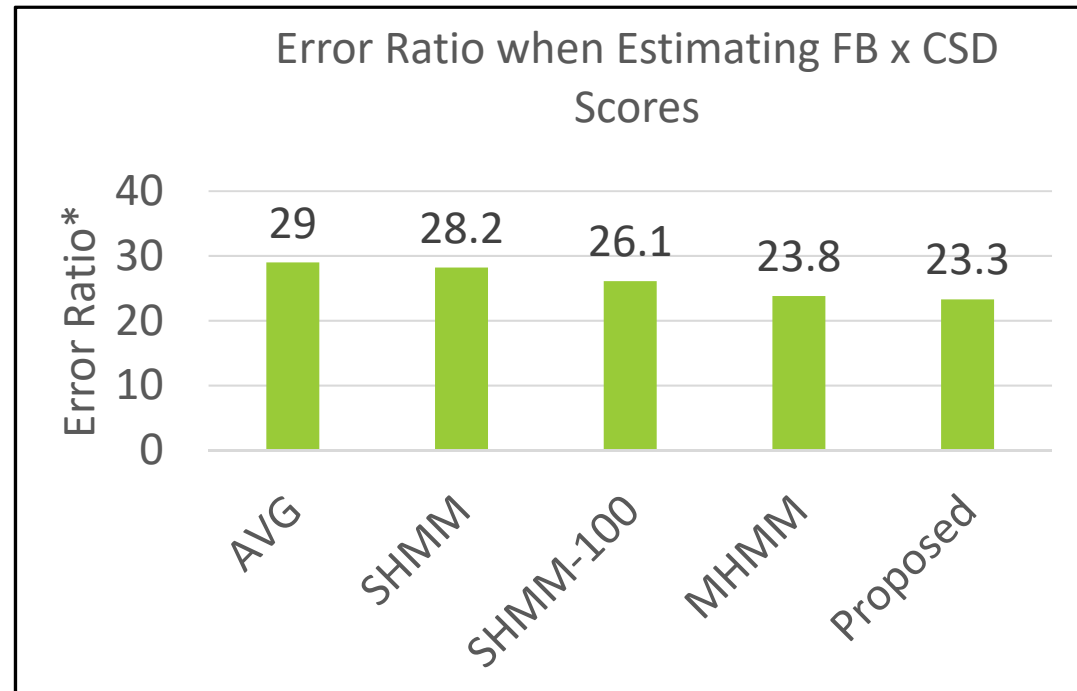


\*Error Ratio = MAE / Total score

# Score Estimation Evaluation: FB x CSD Architecture



Estimated six scores (4-point scale), corresponding to each of the three evaluation criteria for both the front teeth and back teeth.

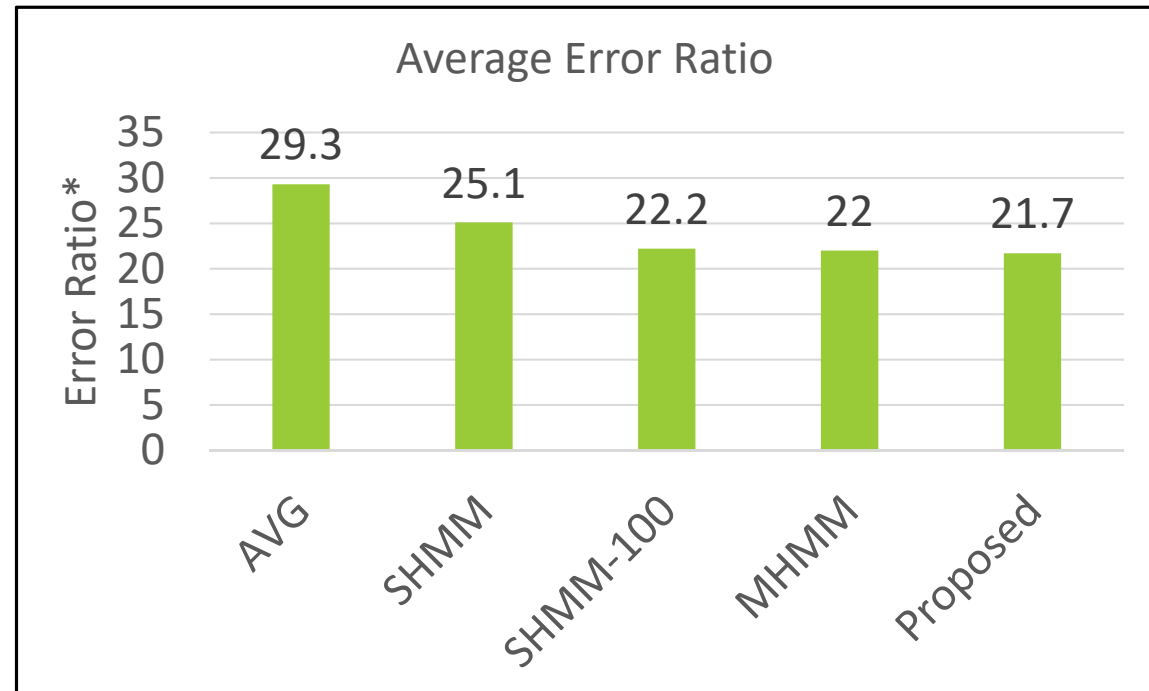


\*Error Ratio = MAE / Total score

# Score Estimation Evaluation: Average Results

Average results for all architectures

1. **Total** (24-pt scale): Estimates one score
2. **FB** (12-pt scale): Two scores: **front** and **back** teeth
3. **CSD** (8-pt scale): Three scores: **coverage**, **stroke**, and **duration**
4. **IO x FB** (6-pt scale): Four scores: One for each area of the mouth **outer front**, **inner front**, **outer back**, and **inner back**
5. **FB x CSD** (4-pt scale): Six scores: **CSD** scores for front teeth and back teeth
6. **IO x FB x CSD** (2-pt scale): Twelve scores: **CSD** for four areas of mouth



\*Error Ratio = MAE / Total score

# Conclusion

---

Proposed a method for evaluating tooth brushing based on audio data

- Create training data using video-based evaluation
  - Enables creation of large amounts of training data
- Perform evaluation on test data using audio-based evaluation
  - Makes method easily accessible to average user
- Tailor HMM sets to score being evaluated
  - Improves performance by avoiding unnecessary distinctions