D4.3 Advanced Fusion System:

Coordinated by UNIS/IDIAP



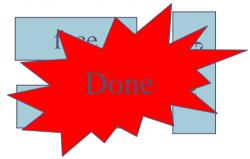
Centre for Vision, Speech and Signal Processing (CVSSP), School of Electronics and Physical Sciences, University of Surrey, U.K.



Overview

Baseline non-adaptive fusion

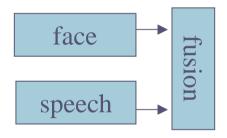
Baseline fusion



D4.1 (m15) & 4.2 (m16)

Advanced non-adaptive fusion

Advanced fusion

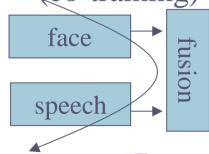


D4.3 (m24) & 4.4 (m30)

Baseline adaptive fusion



Advanced adaptive fusion (co-training)



co-train

D4.5 (m17) & 4.6 (m20)

D4.7 (m26) & 4.8 (m30)

® Dr. Norman Poh except where acknowledged

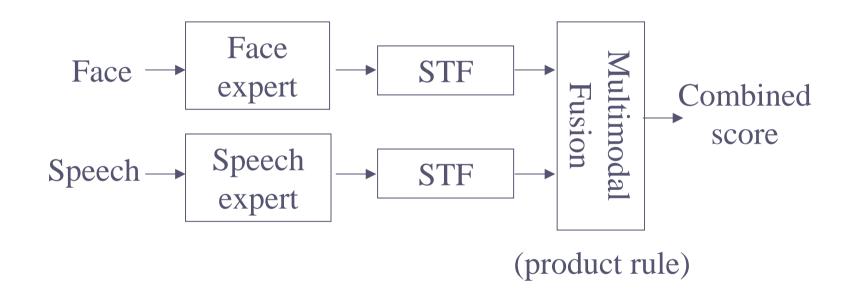
Slide 2

D4.3 Advanced fusion system

- **UNIS**: to investigate frame-level score-level fusion
- IDIAP: to investigage feature-level fusion
- Important dates
 - System deliverable D4.3 (m24)
 - Report deliverbale D4.4 (m30)



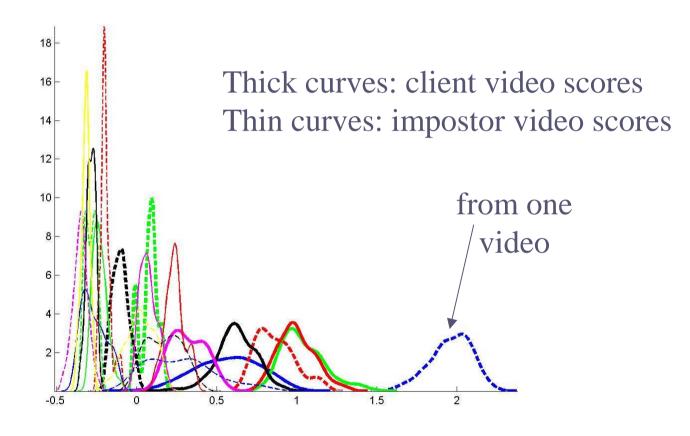
Frame-leve score: Short-term Temporal Fusion (STF)



Note: No interaction between STF of different modalities (a possible future work)

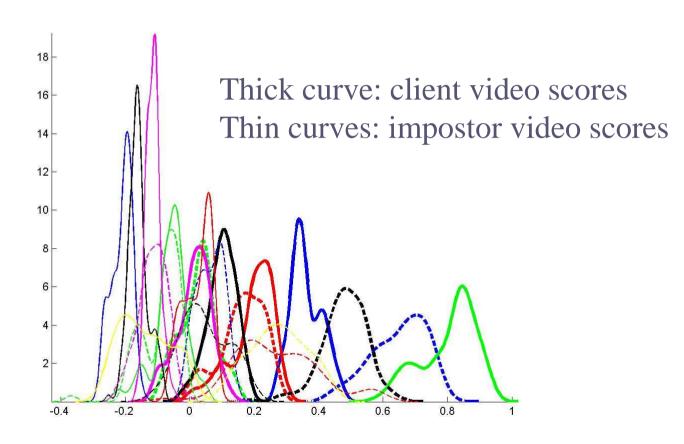


Case Study: Client 1001



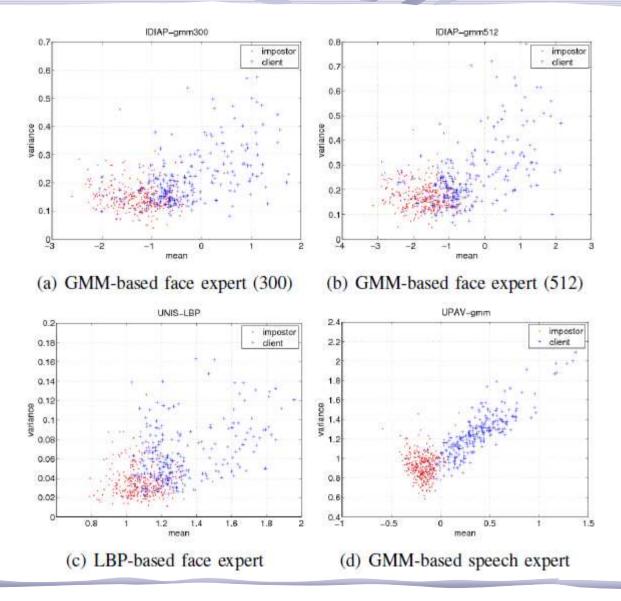


Case Study: Client 1013





Mean-std Scatter Plots



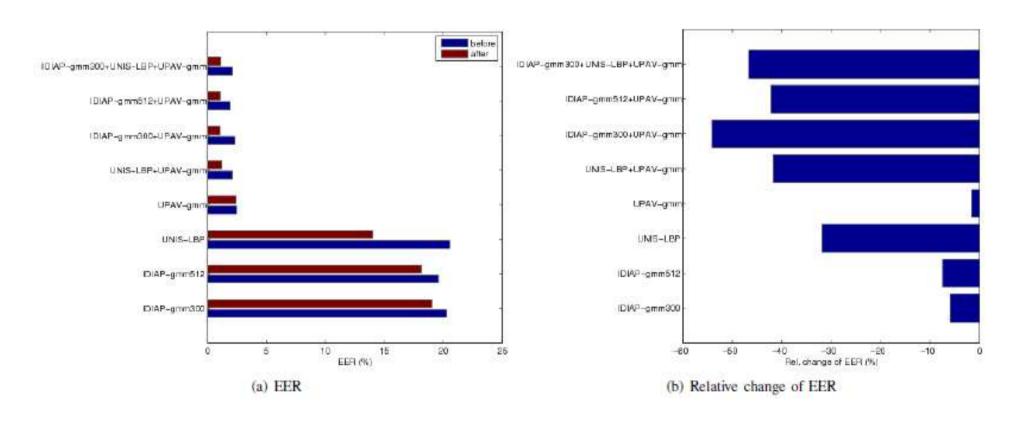


Our Approach

- Based on the premise:
 - Client scores have higher variability than impostor scores
 - The characteristics of video-based score distributions contain some discriminative information
- Train a logistic regression with [mu, sigma, skew, Q1,Q2, Q3, min, max] as features

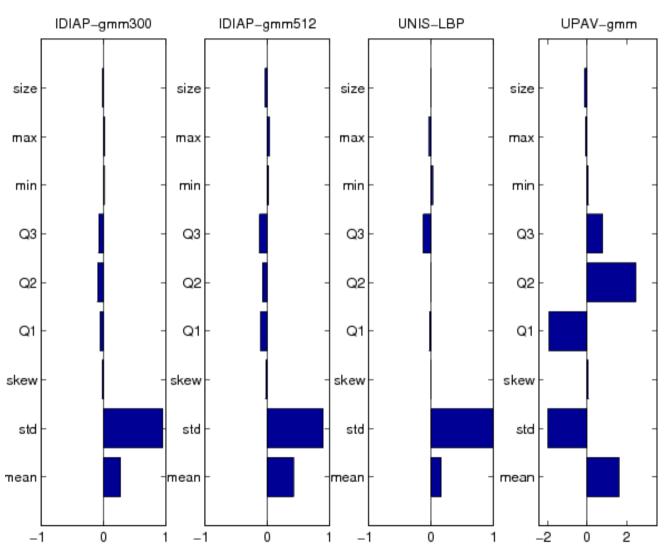


Performance Improvement



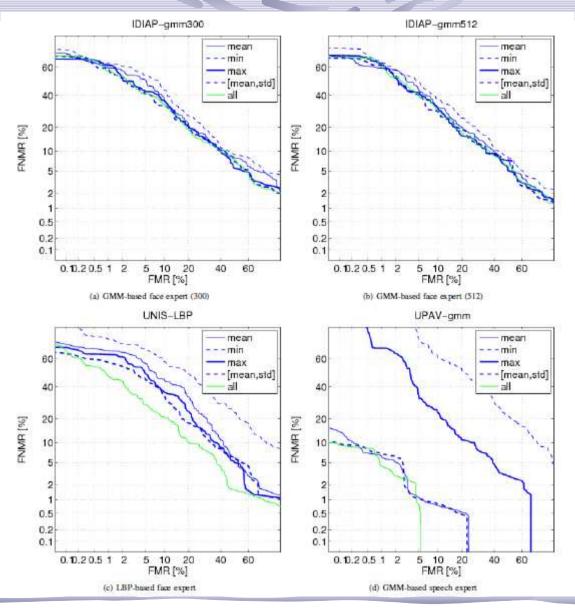


Logistic regression weights



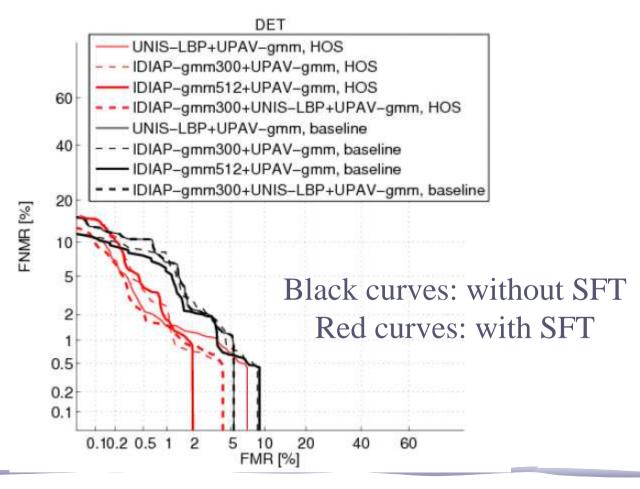


DET curves before/after SFT





DET curves for fusion with/with out SFT





Conclusions

- STF improves the individual expert performance
 - The UNIS face-based LBP expert benefits the most from the STF
 - GMM-based experts benefit only marginally
- In fusion, the relative improvement is still remarkable
- Why larger client variance?
 - Requires a deeper investigation we only have conjecture at this moment – need an experimental simulation

