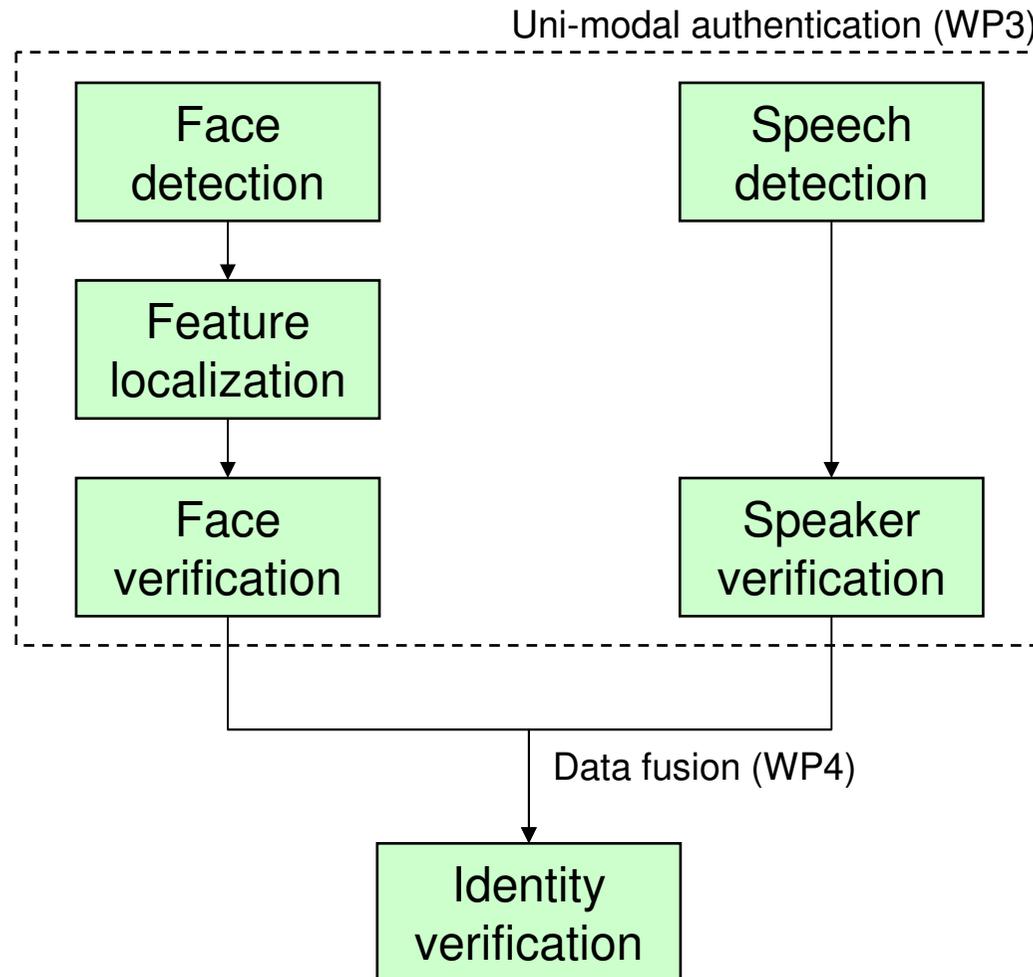


WP3: Uni-modal Segmentation and Authentication

University of Manchester

WP3 Summary



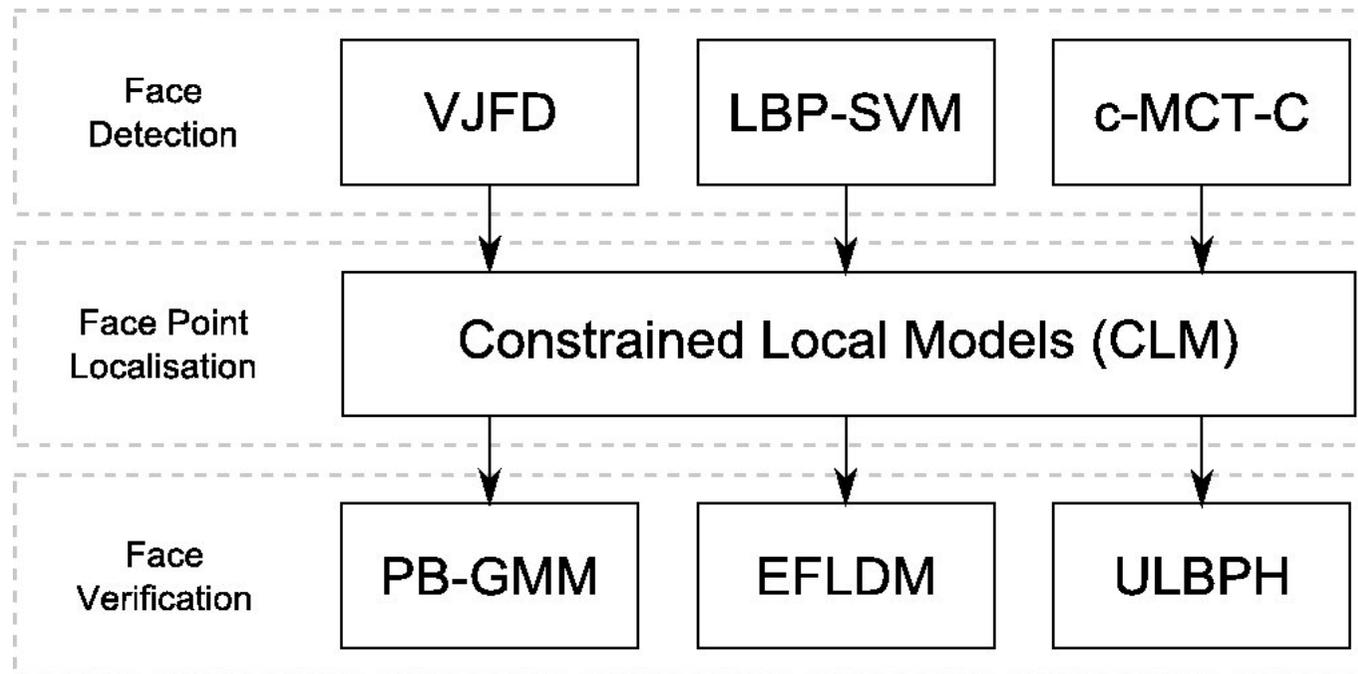
WP3 Summary

- Face authentication
 - Face detection (IDIAP/UOULU)
 - Feature localization (UMAN)
 - Face verification (IDIAP/UNIS)
- Speech authentication
 - Voice activity detection (BUT/LIA)
 - Speaker verification (BUT/LIA)

WP3 Summary

- Deliverables
 - D3.1 (m8) Baseline system software (UMAN)
 - D3.2 (m12) Report on description and evaluation of baseline systems (UMAN)
 - D3.3 (m18) Advanced system software (UOULU)
 - D3.4 (m22) Report on description and evaluation of advanced systems (UOULU)

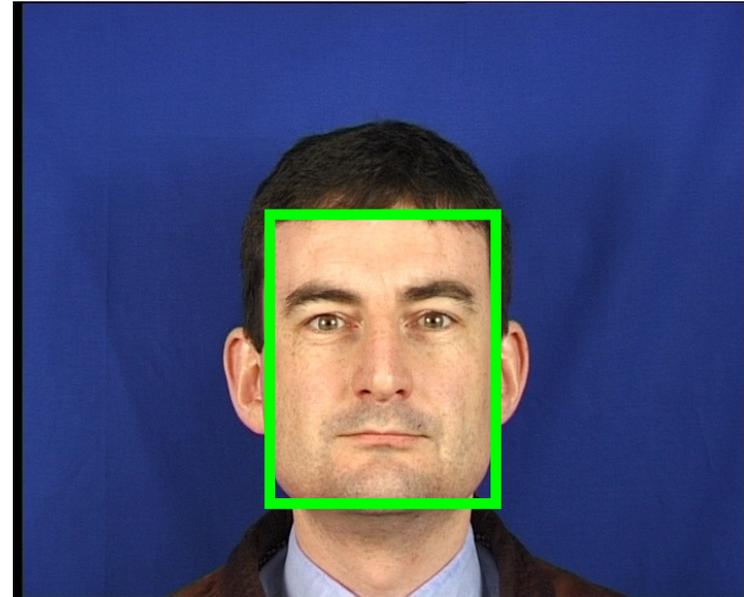
Baseline face authentication



- 1) VJFD- Viola-Jones Face Detector (UOULU)
- 2) LBP-SVM – Local Binary Pattern with Support Vector Machines (UOULU)
- 3) c-MCT-c – Cascade of Modified Census Transform classifiers (IDIAP)
- 4) CLM – Constrained Local Model (UMAN)
- 5) PB-GMM – Parts based Gaussian Mixture Model (IDIAP)
- 6) EFLDM – Enhanced Fisher Linear Discriminant Model (UNIS)
- 7) ULBPH – Uniform Local Binary Pattern Histogram (UNIS)

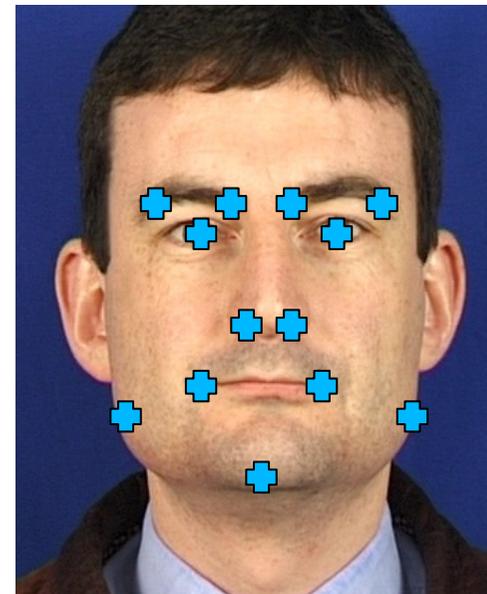
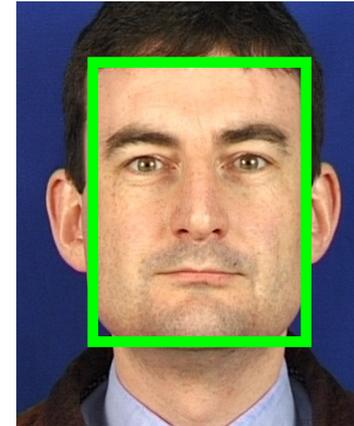
Face detection

- Given an image containing a face, recover its bounding box
- We can then crop out just the relevant information



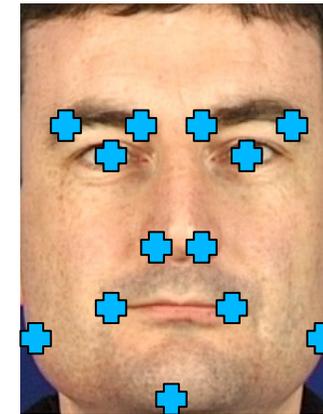
Feature localization

- Given the cropped image of just a face, find the locations of individual features
 - Eyes, nose, mouth, ...
- We can then transform to a normalized co-ordinate frame
 - Remove expression and head rotation



Face verification

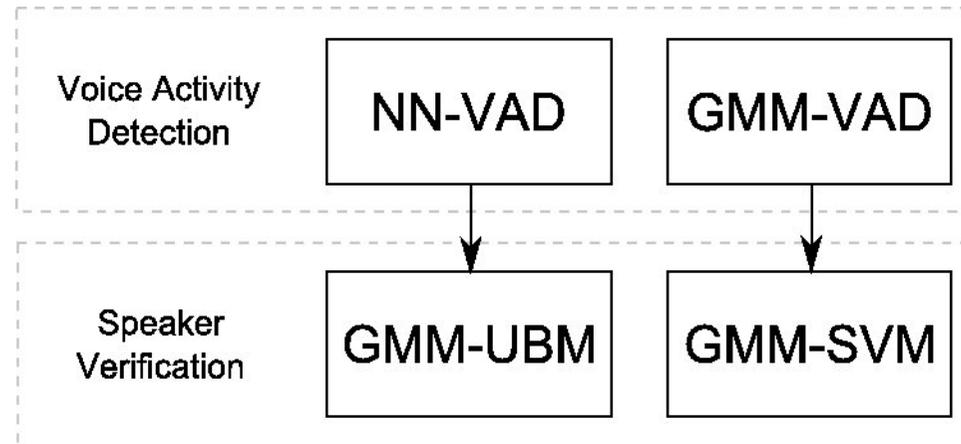
- Given the normalized image of a face and a model of the claimed identity, determine whether the user is who they claim to be



Verifier

Accept/Reject

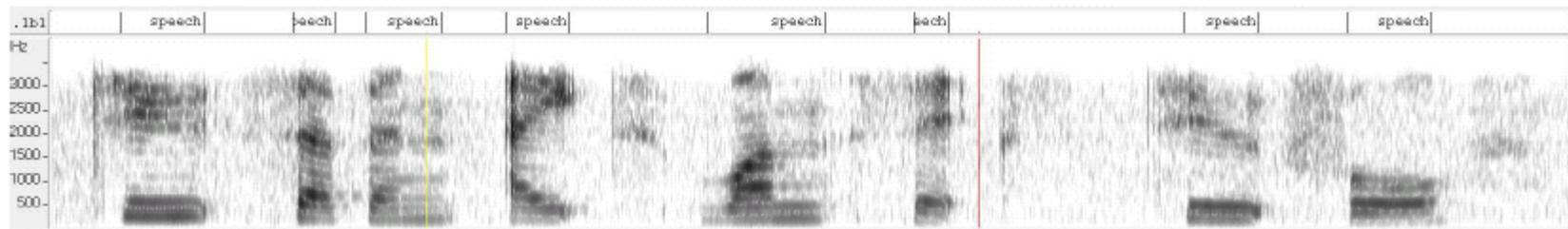
Baseline speech authentication



- 1) NN-VAD – Neural networks based Voice Activity Detection (BUT)
- 2) GMM-VAD – Gaussian Mixture Model based Voice Activity Detection (LIA)
- 3) GMM-UBM – Gaussian Mixture Model and Universal Background Model Paradigm (BUT)
- 4) GMM-SVM – Gaussian Mixture Model and Support Vector Machines (LIA)

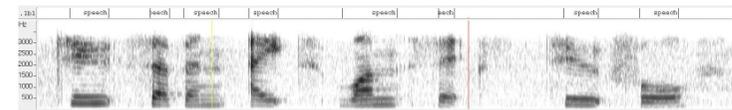
Voice activity detection

- A given speech recording will consist of both speech (signal) and silence (noise)
- VAD determines what is what within the recording and throws away the useless bits



Speaker verification

- Given some speech and a model of the claimed identity, determine if the user is who they say they are
- Likelihood ratio
 - Does the user sound significantly more like X than the average man?



Verifier

Accept/Reject

Project status

D3.1 – Baseline system software

- Software tools completed for baseline algorithms
 - Face detection, feature localization and verification
 - Speech detection and verification
- Problems
 - Nothing from LIA in SVN repository
 - Sparse documentation in places
 - File-naming errors
- Questions
 - Source/binary files in repository?

D3.2 – Baseline system report

- Report completed before Christmas
- Problems
 - Datasets not always consistent (BANCA or NIST?)
- Questions
 - Writing guidelines (e.g. active/passive voice) needed?
 - Best way to implement all detectors and all verifiers on all datasets?

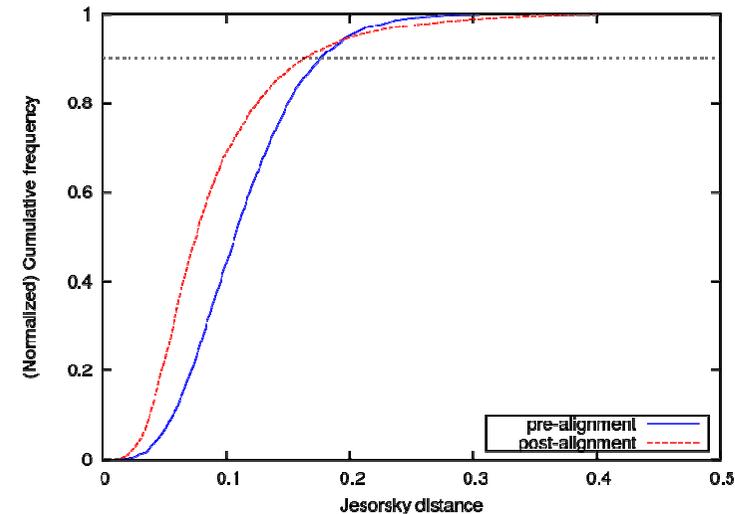
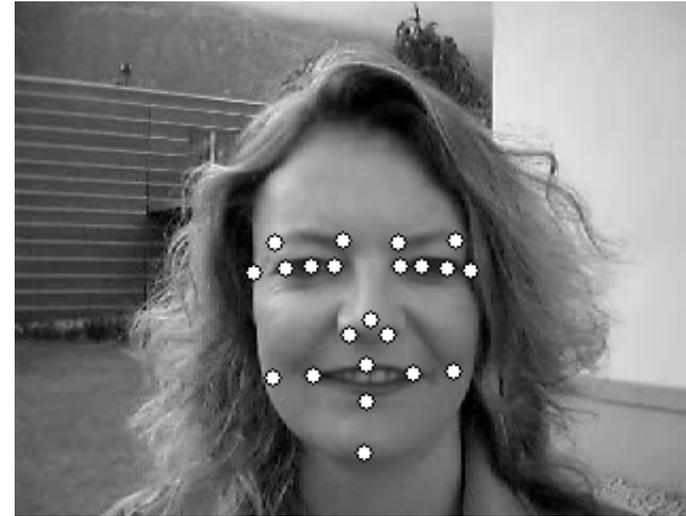
Project progress (UMAN)

Summary

- Phil Tresadern joined August 2008 (5 months)
- Harish Bhaskar joined September 2008 (4 months)
- Contributions to deliverable D3.1
- Co-ordinated deliverable D3.2
- Commenced research towards deliverable D3.3

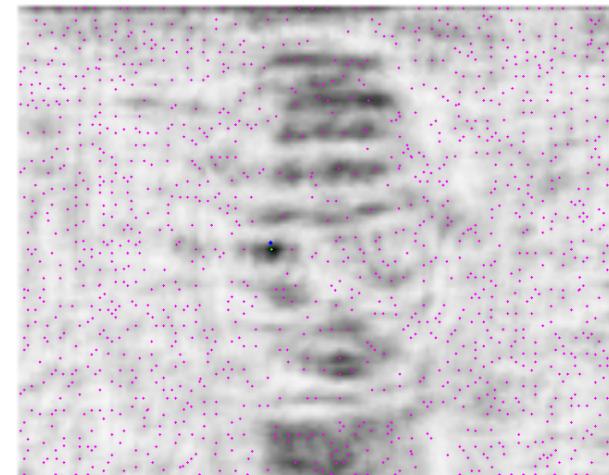
Work completed

- Produced baseline system software for facial feature localization (D3.1)
- Co-ordinated deliverable D3.2
- 2 data collection sessions completed (WP2)

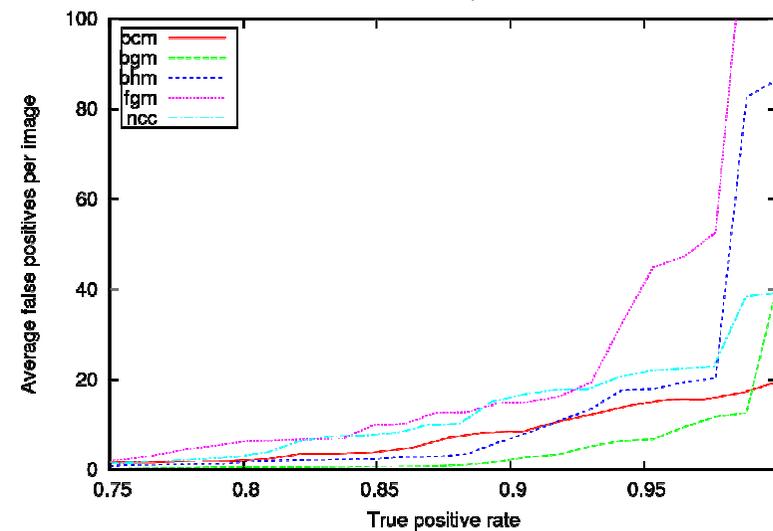


Work ongoing

- Improving facial feature detection (WP3)
 - Comparison of different appearance models for detecting the right eye as a preliminary example

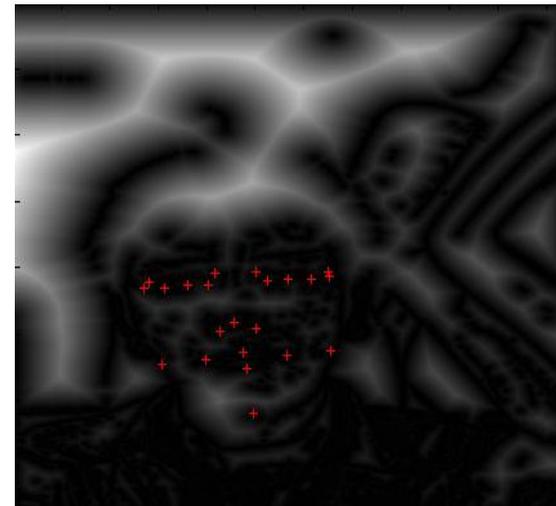


radius = 10 pixels



Work ongoing

- Graph-based face alignment methods (WP3)
 - Represent deformation field from one frame to the next by a grid of discrete displacements
 - Unary costs enforce consistent image-image matching
 - Pairwise costs enforce spatial priors



Work ongoing

- Graph-based face alignment methods (WP3)
 - Markov Random Field solvers (e.g. TRW-S) used to minimize error function



Future work

- Extend comparison of facial feature detectors (D3.3)
 - Include more examples (e.g. AdaBoost)
 - Develop most promising models to improve accuracy and reduce false positives
- Incorporate improved facial feature models into graph-based localization scheme (D3.3)
- Liaise with UOULU with respect to evaluations of advanced systems (D3.4)

Future work

- Investigate data fusion techniques and model adaptation (WP4)
- Investigate scalability of feature detectors (WP5)
 - Consider implementation for maximum efficiency on low-power, fixed-point processors

Publications

- None submitted to date
- Forthcoming conference deadlines
 - ICCV: 10th March 2009
 - BMVC: 27th April 2009

WP4: Joint Bi-Modal Authentication and Model Adaptation

University of Manchester

WP4 Planning (UMAN)

- Lip-tracking
 - Synchronize audio and video
- Blink detection
 - Liveness detection
 - Eliminating images with eyes closed
- Head pose estimation
 - Frontal faces are easier to authenticate

WP4 Planning (UMAN)

- Investigate the use of group-specific models for improved localization
 - Group subjects based on age, gender etc. and generate group-specific models
 - Select appropriate model using known data about subject
 - Evaluate any improvement in tracking due to group-specific model

WP4 Planning (UMAN)

- Investigate the use of person-specific models for further improved localization
 - Given enough data from a specific person, develop a person-specific model
 - Model could potentially be stored in mobile phone (most phones are used by only one person)

WP4 Planning (UMAN)

- Investigate model adaptation over time
 - Adapt model to track changes in appearance (e.g. growing facial hair)
 - Q: How do we decide when to update model?
- Will environmental effects (e.g. illumination) dominate over true changes in appearance (e.g. make-up, facial hair)

WP5: Scalability

University of Manchester

WP5 Planning (UMAN)

- Appearance models:
 - Investigate their implementation on fixed-point processor. E.g. PCA may be difficult, histograms and Haar wavelets are a bit easier
 - Evaluate number of mathematical operations required for each evaluation. Can we reduce the number of math ops per evaluation and the number of evaluations required (e.g. cascade of classifiers)
 - How to reduce memory requirements? Feature selection methods (e.g. AdaBoost). How critical is it to estimate confidence?

WP5 Planning (UMAN)

- Graph-based matching
 - Currently very complex and not well-suited for fixed-point, low-power processors
 - Can they be adapted to the processors that we will be working with?

WP5 Planning (UMAN)

- Distribution
 - If not, how can we distribute processing between mobile and server? What is the trade-off between sending data to server for processing vs processing on the mobile?
 - What processes do we have in common? E.g. integral image/census transform may be computed for detection then used for localization and verification for free