

# ***Testing the validity and reliability of forensic voice comparison based on reassigned time-frequency representations of Chinese /iau/***

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# Paradigm for evaluation of FVC evidence



- Likelihood-ratio framework:
    - Statement of strength of the evidence as an answer to a specific question
- $$LR = \frac{p(E | H_p)}{p(E | H_d)}$$
- Quantitative measurements, statistical models, databases representative of the relevant population
  - Testing of validity and reliability under conditions reflecting those of the case

- Fulop & Disner (2007, 2009):
  - Pruned T-F-reassigned spectrograms of short vowel segments ([æ], [a] etc.)
  - visual comparison of spectrograms by human experts (“voiceprint”)
  - Fulop (2011): U.S. Patent 8,036,891 B2
- Fulop & Kim (2013):
  - Quantitative approach
  - Automatic SVM-based closed-set identification
  - 24 enrolled speakers, 6 test segments

- Short-time Fourier transform of /iau/
- Channelized Instantaneous Frequency (CIF)
- Local Group Delay (LGD)

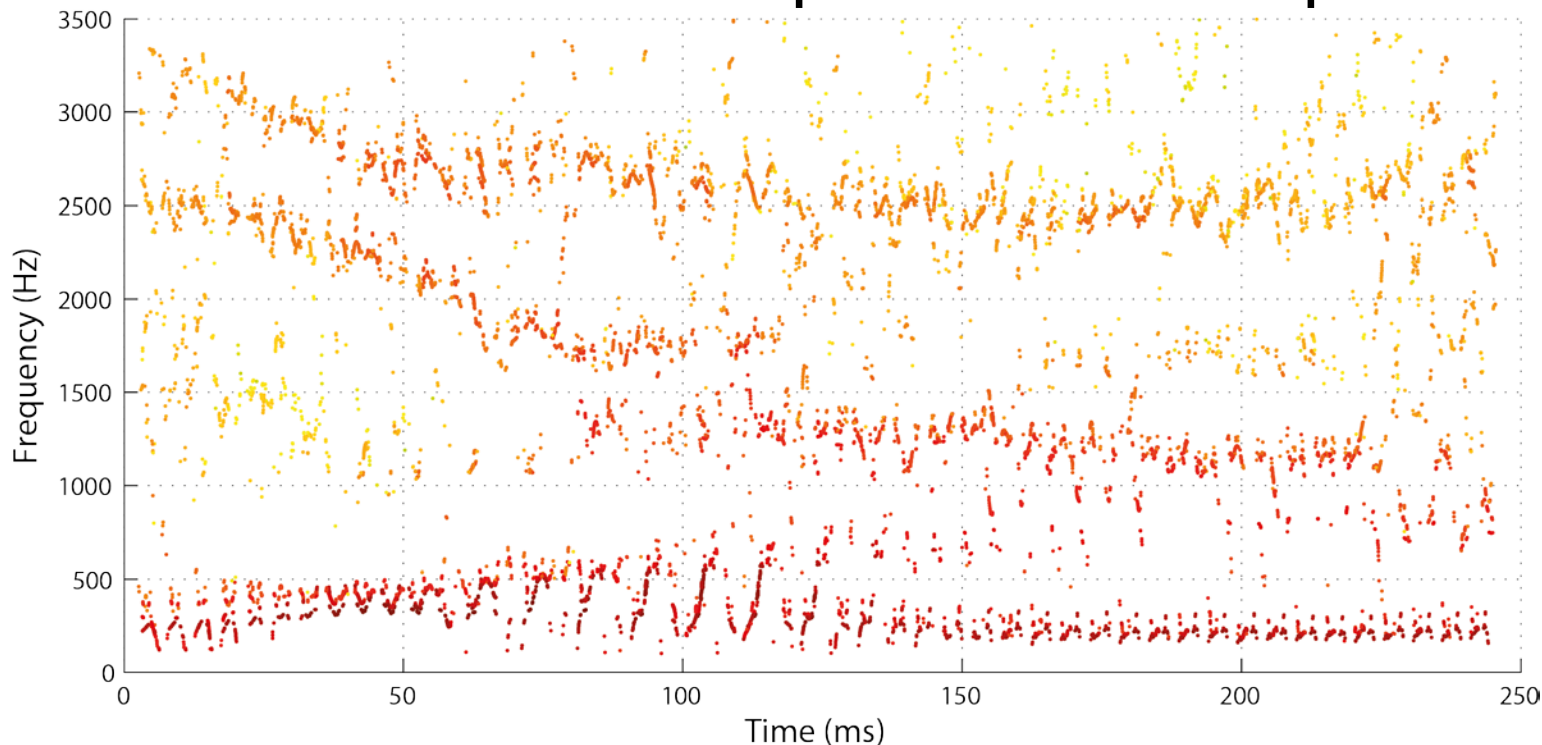
$$\begin{aligned}\text{CIF}(\omega, T) &= \frac{\delta}{\delta T} \arg(X_h(\omega, T)) \\ \text{LGD}(\omega, T) &= \frac{\delta}{\delta \omega} \arg(X_h(\omega, T))\end{aligned}$$

- “Reassign” T-F magnitudes to locations corresponding to local center of gravity

# TF reassigned spectrograms



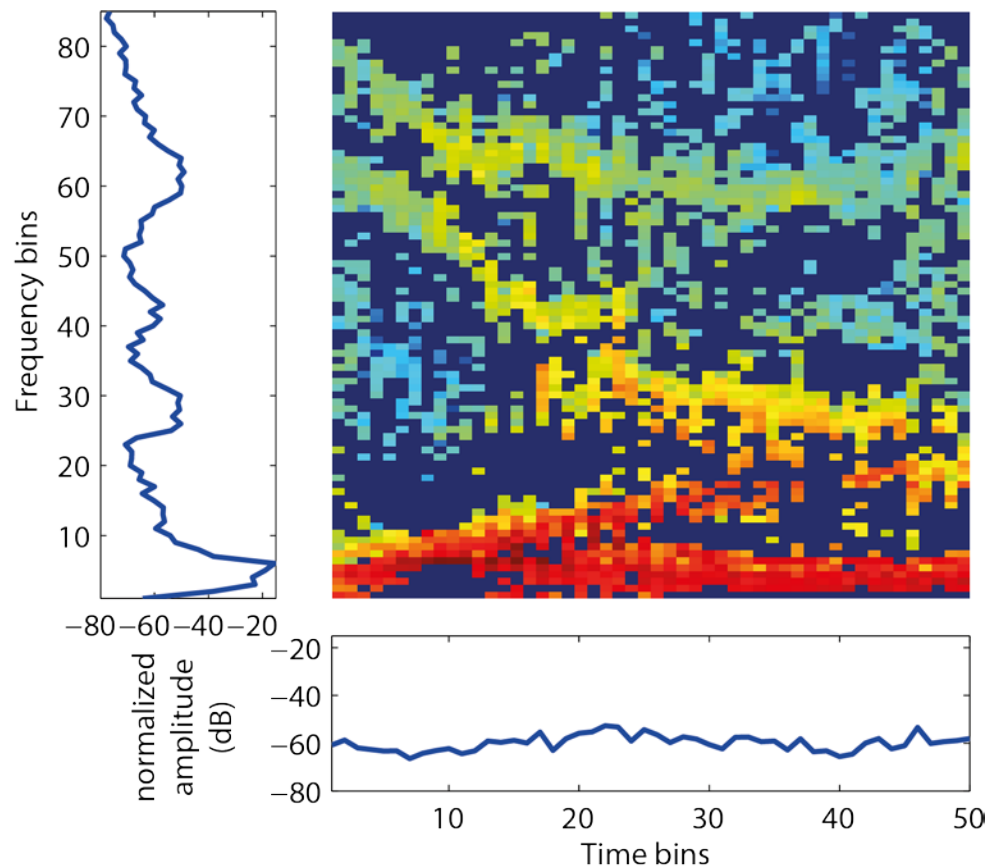
- Pruning (threshold) to reduce noise/artefacts
  - Based on second-order mixed partial derivative (Nelson, 2001)
  - Set to retain line components and impulses



# TF feature representation – TFR AVG



- Fulop & Kim (2013): Feature representation based on discretization using a coarse grid
  - 50 time bins
  - 85 frequency bins
- Dimensionality reduction via PCA
  - 10 time features
  - 20 frequency features

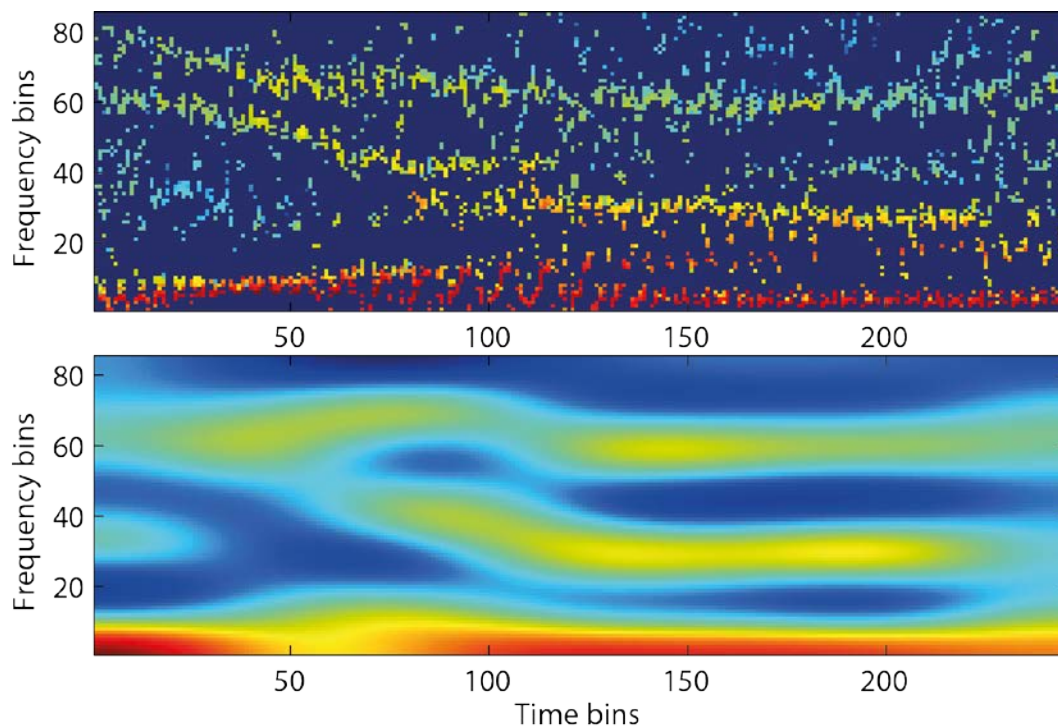




# TF feature representation – TFR DCT



- Chinese /iau/ triphthong:
  - Significant correlation over time and frequency
  - 2D Discrete cosine transform (DCT)

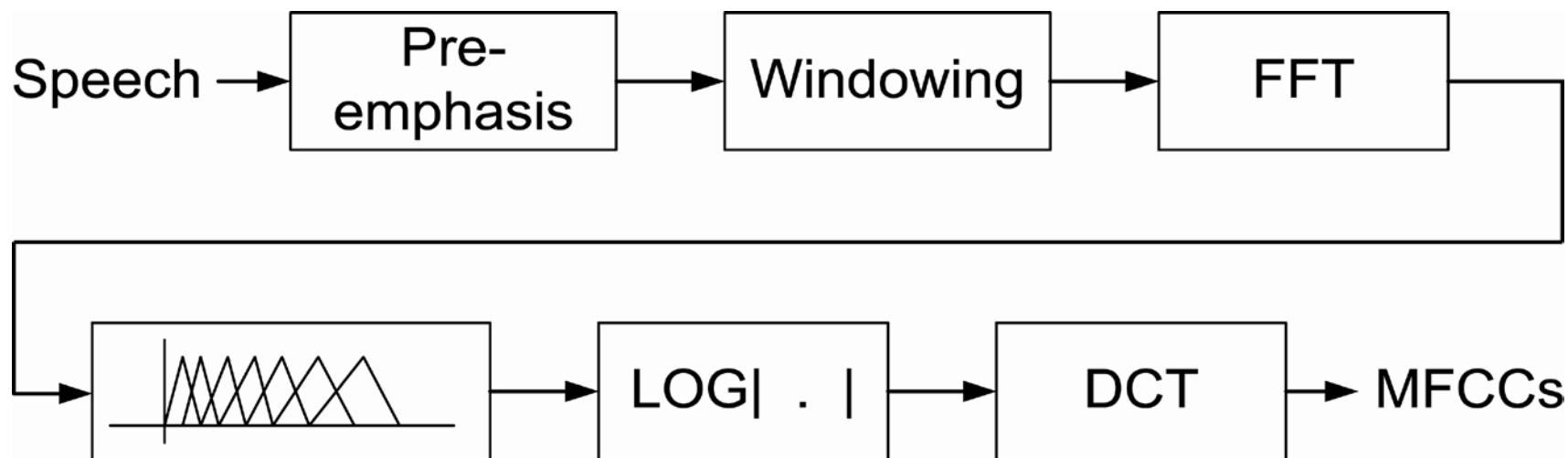


lower-order  $7 \times 7$  coefficients

# Feature representation – MFCC-on-/iau/



- Mel frequency cepstral coefficients (MFCC)
  - Common feature in FVC / speaker recognition
  - Extracted from /iau/ triphthong tokens
  - 16 MFCC + 16 Delta ( $\Delta$ ) coefficients





- Score obtained using Gaussian mixture model-Universal background model (GMM-UBM) approach

$$\lambda = (p_i, \mu_i, \Sigma_i)_{i=1, \dots, M} \quad s = \frac{1}{N} \sum_{j=1}^N \log \left( \frac{p(x_j | \lambda_{\text{suspect}})}{p(x_j | \lambda_{\text{UBM}})} \right)$$

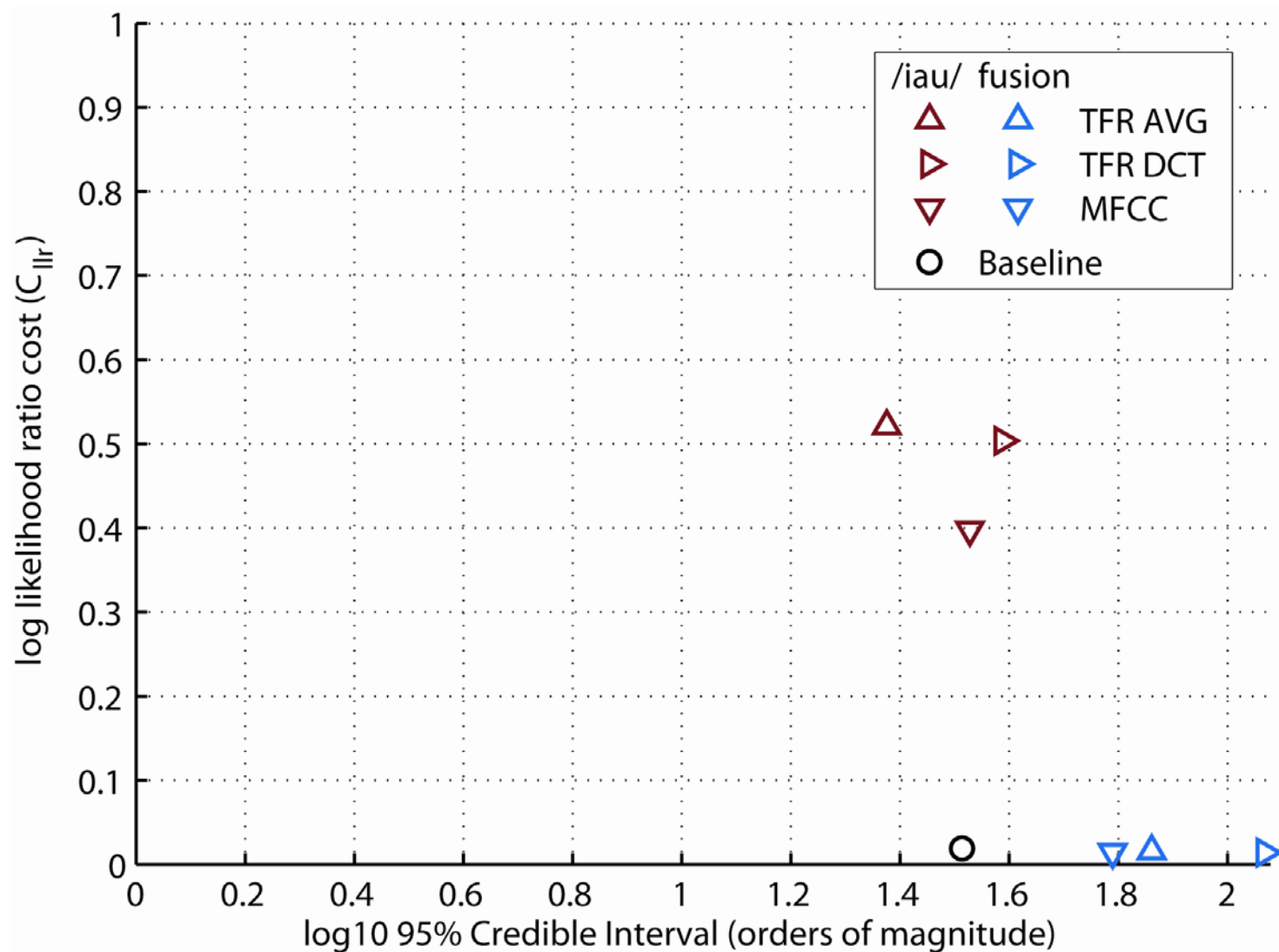
- Logistic regression calibration and fusion
- Baseline automatic FVC system
  - Entire speech-active portion of recording
  - 16 MFCC + 16 delta ( $\Delta$ ) coefficients
  - 1024 Gaussian mixture components (UBM)

- 60 female Standard Chinese speakers
- Split into 3 groups of 20 speakers
  - background set
  - development set
  - test set
- Manually marked /iau/ triphthongs
- Information-exchange task over telephone
- High quality and mobile-to-landline data
- Two recording sessions separated by 2–3 weeks

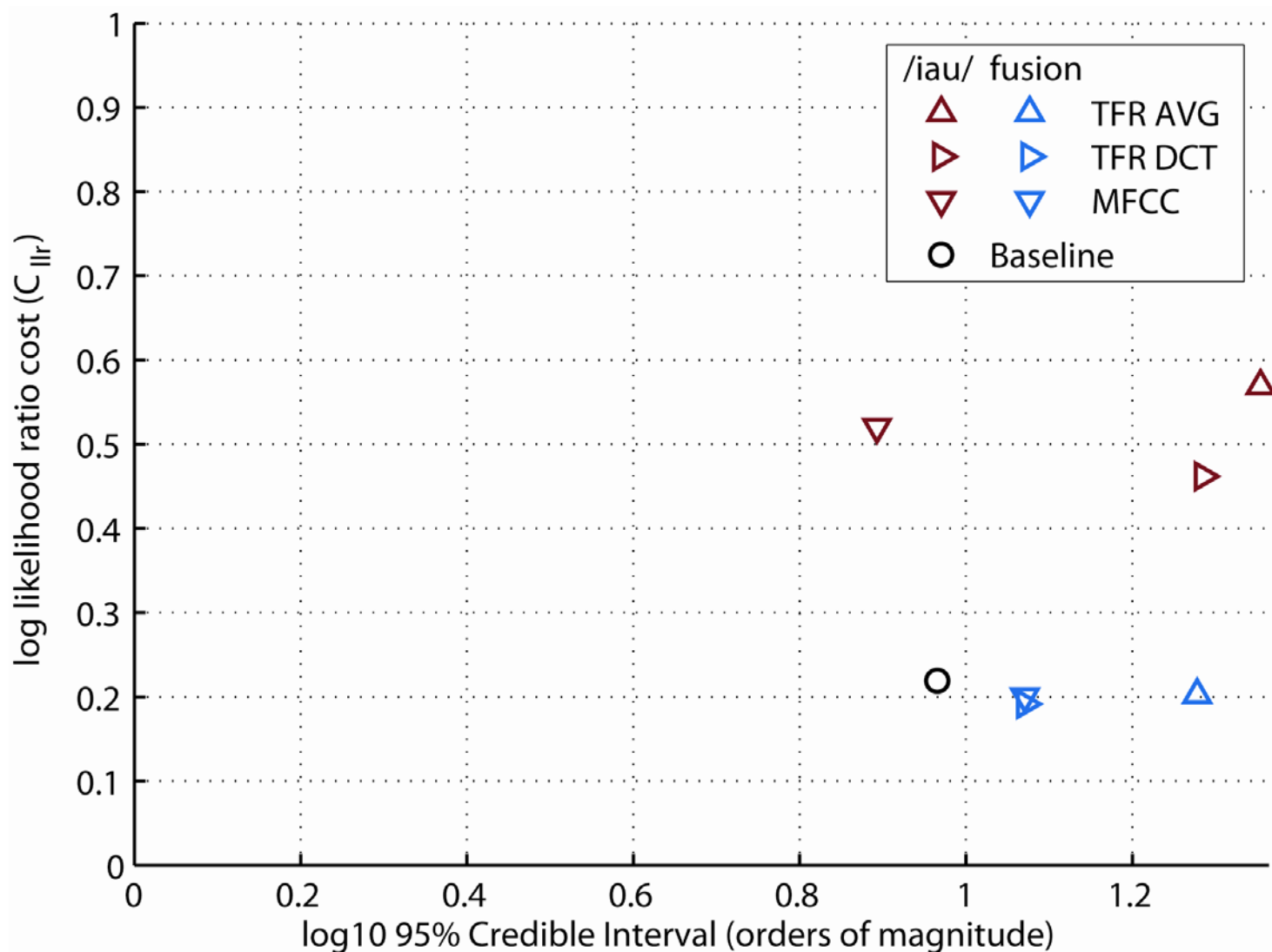
<http://databases.forensic-voice-comparison.net/>

- Validity / Accuracy
  - log-likelihood ratio cost ( $C_{llr}$ ) metric
- Reliability / Precision
  - 95% credible interval (Morrison, 2011)
- Conditions:
  - High-quality v high-quality
  - Mobile-to-landline v mobile-to-landline
  - High-quality v mobile-to-landline

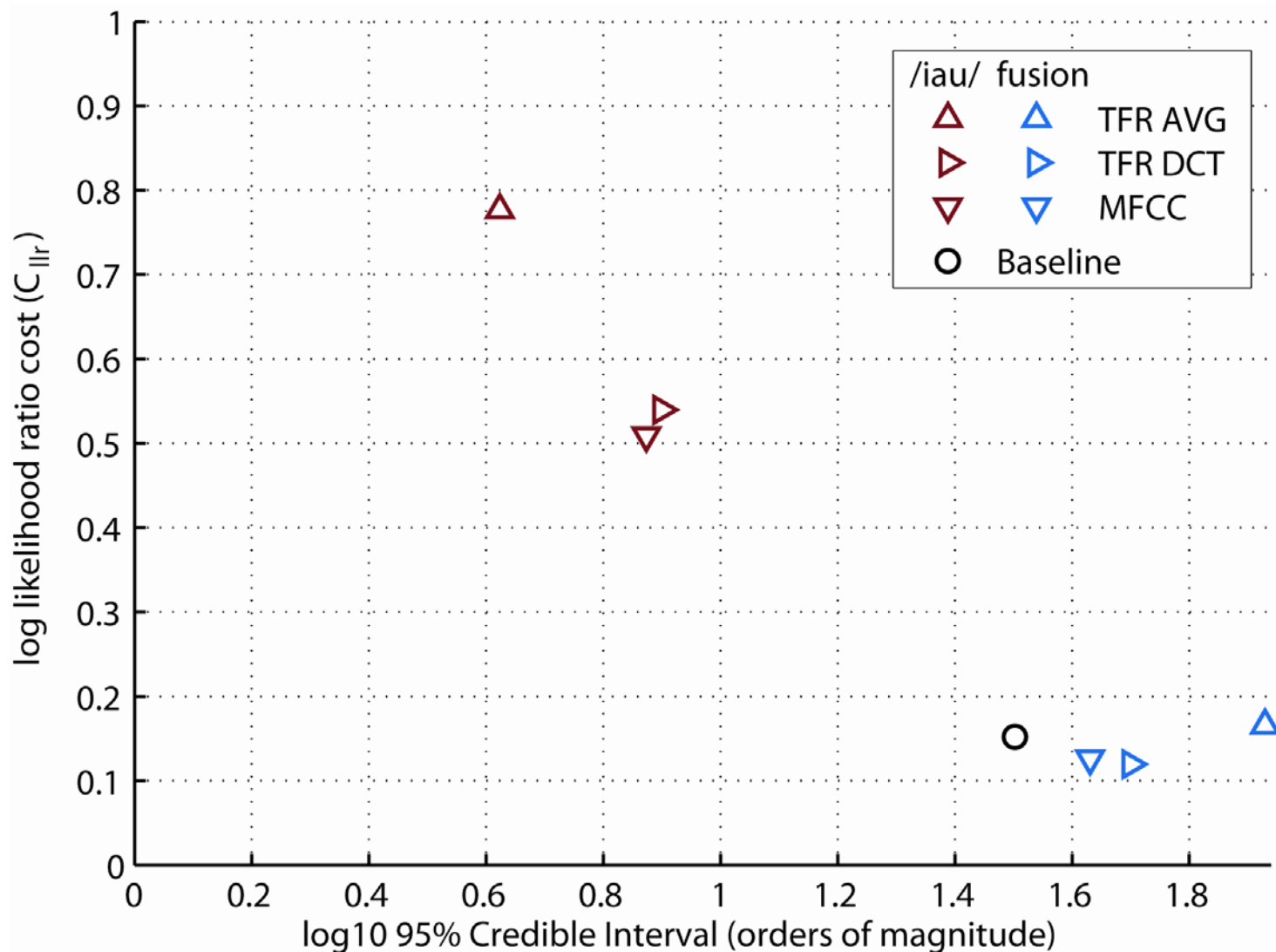
# Results – high-quality v high-quality



# Results – mobile-to-landline v mobile-to-land

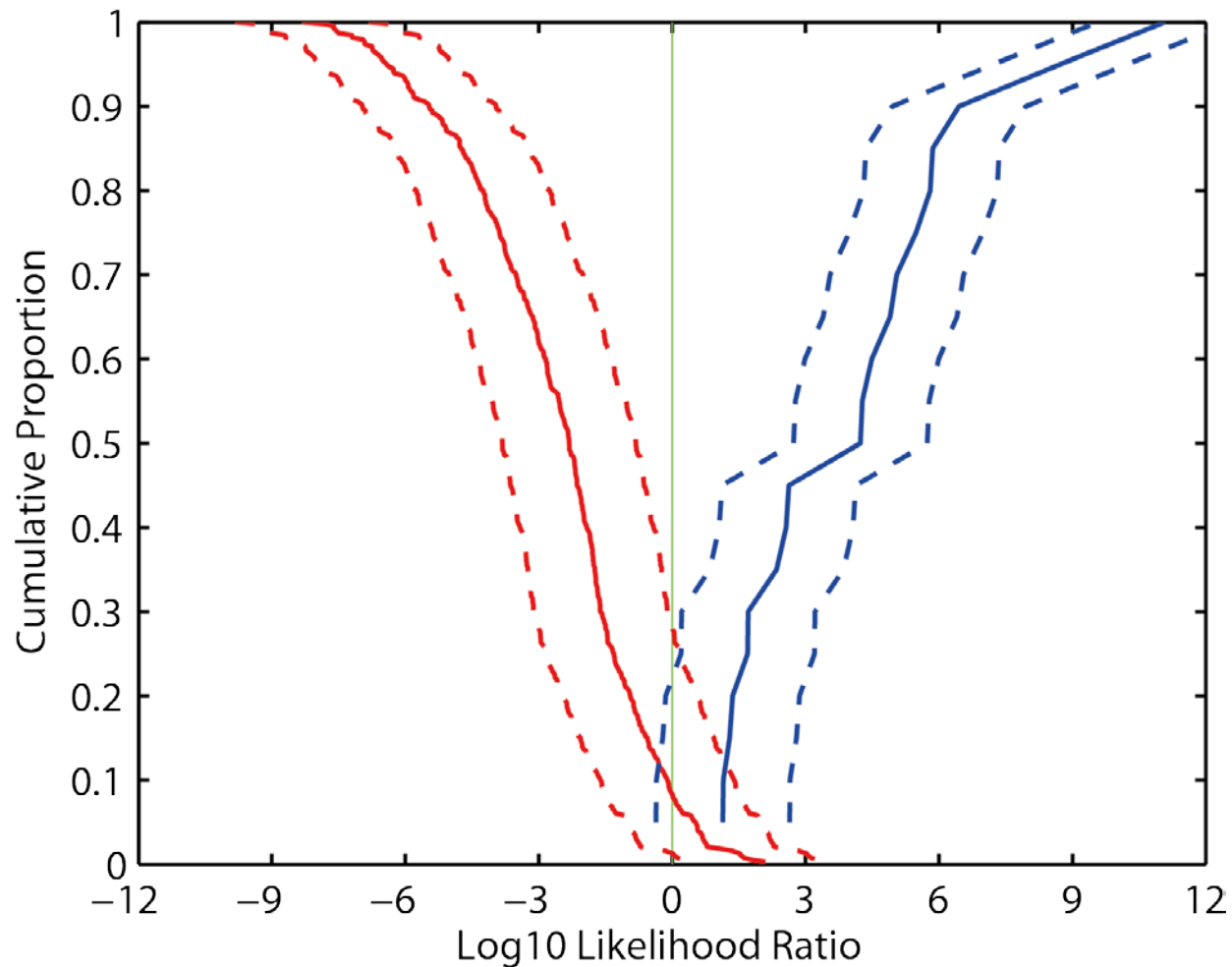


# Results – mobile-to-landline v high-quality

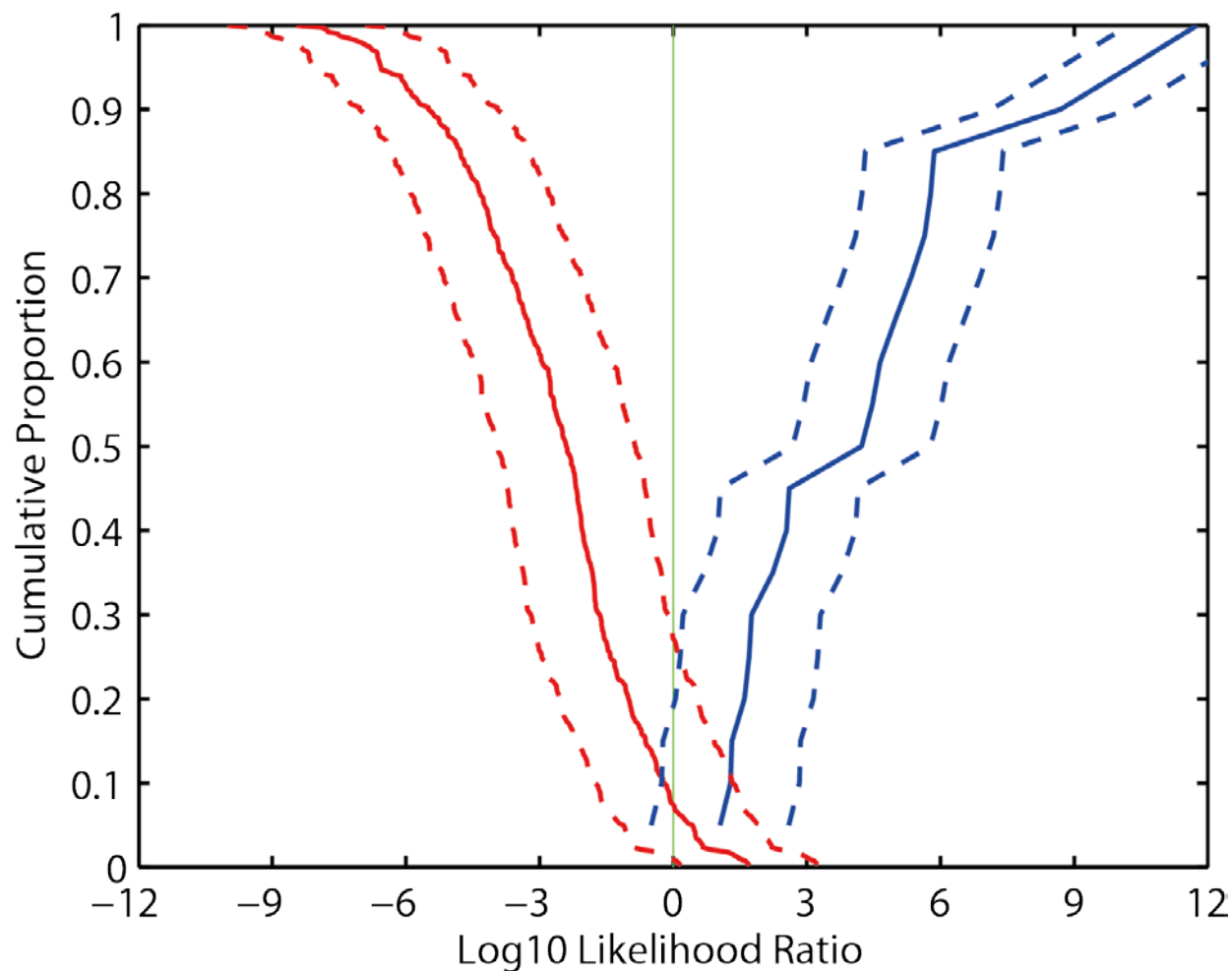




# Tippett plot – Baseline system



# Tippett plot – Fusion Baseline + TFR DCT



# Conclusion



- High-quality v high-quality
  - no substantial improvement
- Mobile v mobile, mobile v high-quality
  - Improvement in validity, reliability deteriorates
  - TFR DCT improves upon TFR AVG
  - MFCC-on-/iau/ similar or slightly better
- Caveat:
  - Results give only an indication of performance (not tested: background noise, reverberation, ..)
  - Testing on per-case basis



Thank You!!

# References



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