Speaker Detection by the Individual Listener and the Crowd: Parametric Models Applicable to Bonafide and Deepfake Speech





Tomi Kinnunen¹, Rosa González Hautamäki^{2,1} Xin Wang³, Junichi Yamagishi³

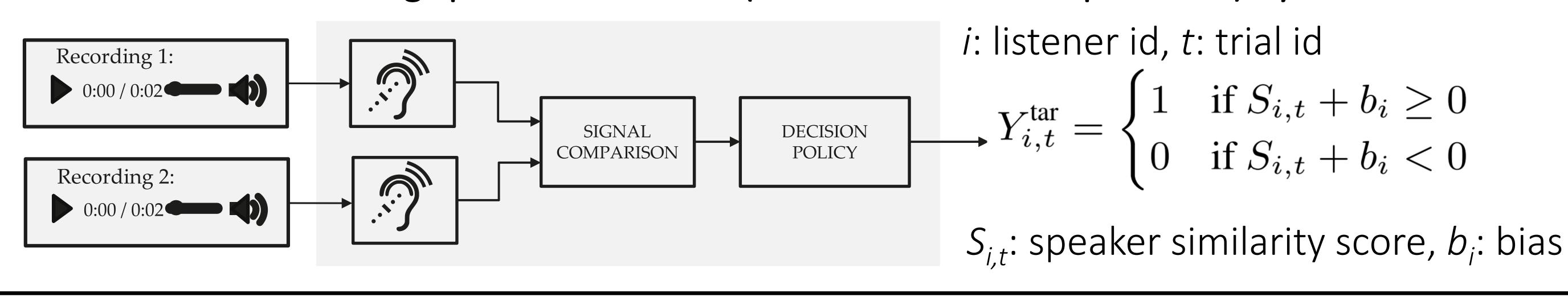




¹University of Eastern Finland, ²University of Oulu, ³National Institute of Informatics

Speaker recognition by humans is relevant to (1) neurosciences, (2) human-in-the-loop applications, (3) evaluation of speech synthesis. Often data is analyzed using nonparametric methods. Our work seeks to remind about the half-forgotten **signal detection theory** for parameteric modeling of speaker detection by (groups of) listeners, demonstrated through two case studies.

Modeling speaker detection ("same or different speaker?") by listeners



A generalized linear mixed effects (GLME) model

"The Crowd" part: generative between-listener model of the detection model parameters

$$\begin{bmatrix} b_i \\ d_i \end{bmatrix} \sim \mathcal{N} \left(\begin{bmatrix} b + b^{(k_i)} \\ d + d^{(k_i)} \end{bmatrix}, \begin{bmatrix} \sigma_u^2 & \rho \sigma_u \sigma_v \\ \rho \sigma_u \sigma_v & \sigma_v^2 \end{bmatrix} \right) \qquad \begin{matrix} k_i & \text{: group id} \\ \sigma_u^2 & \sigma_v^2 & \text{: between-listener variances of } b \text{ and } d \\ \rho & \text{: correlation of } b \text{ and } d \end{matrix}$$

"The individual listener" part: parametric detection model for listener i

$$S_{i,t} = d_i X_t^{\text{tar}} + \varepsilon_{i,t}$$

Miss and false alarm rates (under logistic model)

 d_i : discrimination parameter

 $P_{\text{miss}}^{(i)} = 1 - \Pr(Y_{i,t} = 1 | X_t^{\text{tar}} = 1) = (1 + e^{b_i + d_i})^{-1}$

 X_t^{tar} : ground-truth (0 or 1)

 $P_{\text{fa}}^{(i)} = \Pr(Y_{i,t} = 1 | X_t^{\text{tar}} = 0) = (1 + e^{-b_i})^{-1},$

 $\varepsilon_{i,t}$: residual (here, either normal or logistic)

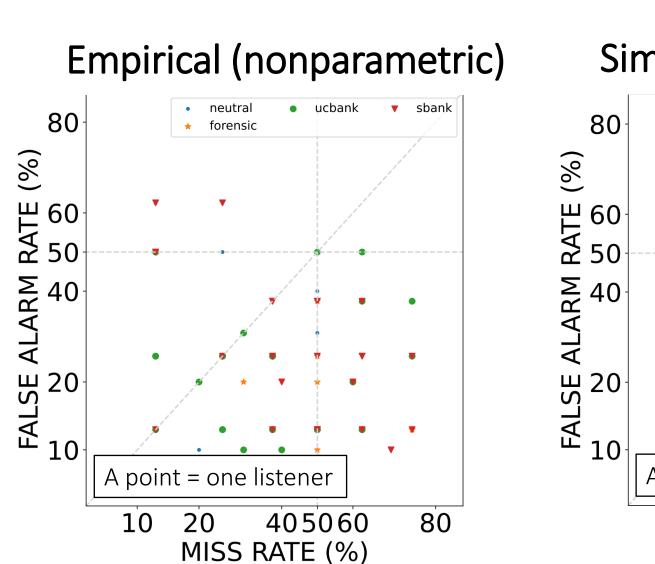
Case study I: impact of role-play

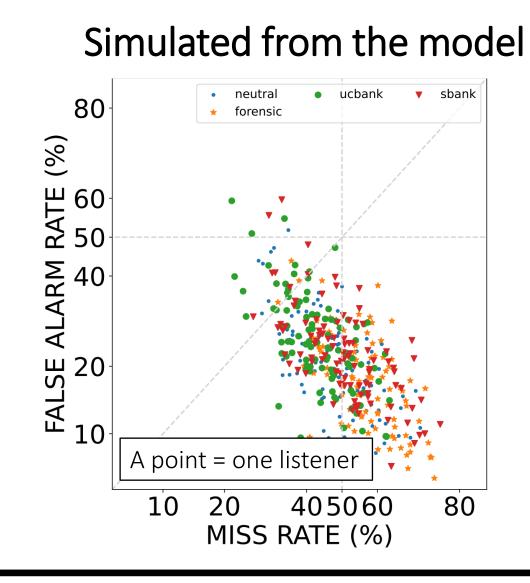
100 target + 100 nontarget trials from VoxCeleb1-E, AMT experiment: 200 trials x 5 repetitions of each x 4 groups = 4000 ratings, 60 listeners (disjoint sets) per panel ($4 \times 60 = 240$ listeners)

	Probit		Logit	
Scenario	$b+b^{(k_i)}$	$d+d^{(k_i)}$	$b+b^{(k_i)}$	$d+d^{(k_i)}$
Neutral	-1.352	1.484	-2.349	2.558
Forensic	-1.563	1.258	-2.757	2.274
User-c. bank	-1.149	1.407	-1.970	2.390
Secure bank	-1.321	1.230	-2.272	2.126
	0.00		0 7 7	

 $\rho = -0.66$

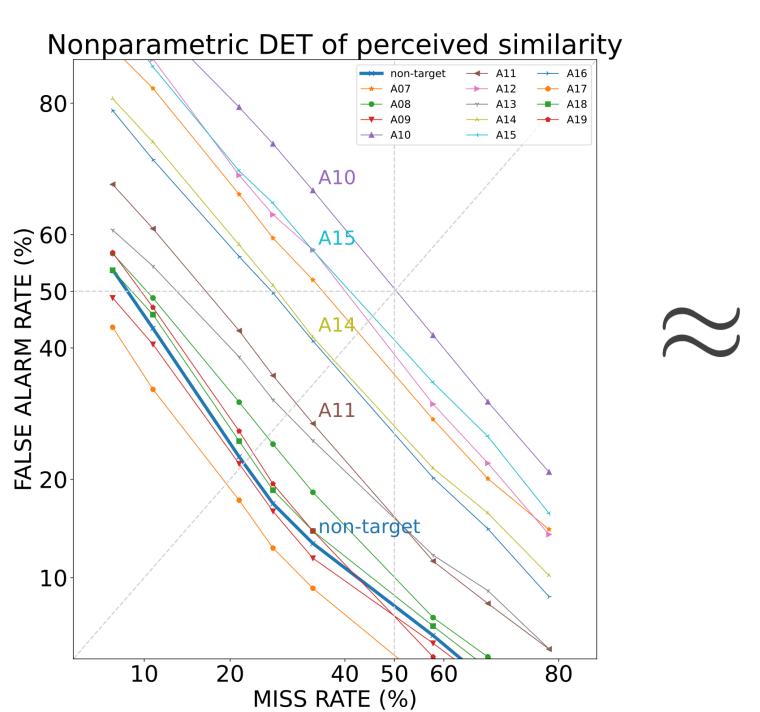
$$\rho = -0.75$$



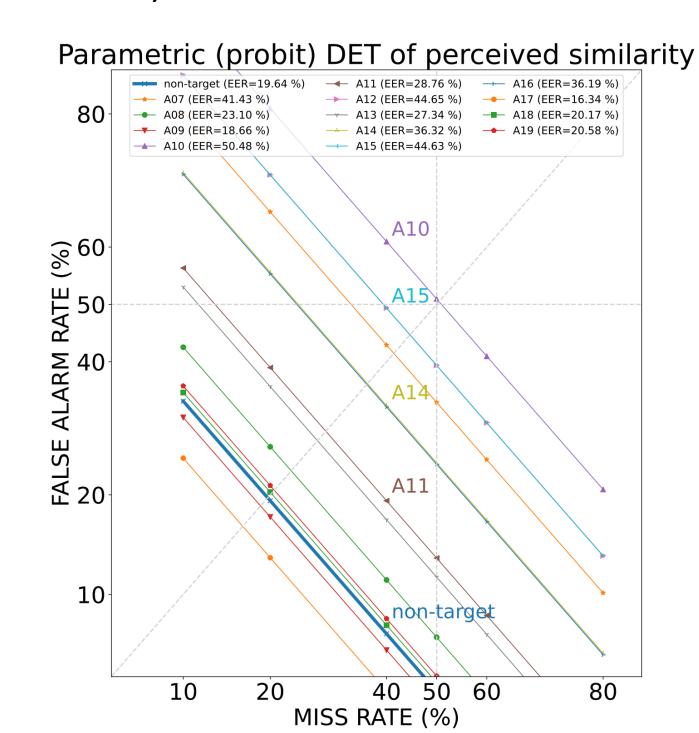


Case study II: potential to simplify similarity rating scale

Speaker similarity ratings in ASVspoof 2019 data (deepfake / spoofed speaker identities). 13 different spoofing methods, rated by 1145 listeners using a 10-point scales \rightarrow converted to binary decisions for our model $(1...5 \rightarrow 0; 6..10 \rightarrow 1)$



Nonparametric DETs based on 10point rating scale



Parametric DETs fitted to binarized listener responses.