# Cluster analysis of voice quality ratings: Identifying groups of perceptually similar speakers

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#### INTRODUCTION

Investigations of voice similarity  $\rightarrow$  increasingly important in various fields:

- 1 Voice casting (Obin & Roebe 2016)
- To determine the set of target actors that are the most similar to the speech recording of a source actor
- E.g. to transfer a film or video game from a source language to a target language with few voices available for each language
- 2 Voice parades (forensic application)
- Perceptual equivalent of a visual identification line-up
- Used when a witness to a crime could not see the perpetrator's face but could hear him/her speak
- Design by forensic phoneticians:
  - to ensure that ear witness evidence is conducted fairly, the forensic expert chooses a set of similar voices to the suspect (foils)
  - earwitnesses are then asked if they can recognize the offender's voice from the selection of voices (de Jong et al. 2015)
  - undesirable effects: suspect's voice stands out because it is not similar enough to the foils and it is too easy for the earwitness to pick out  $\rightarrow$  test is unfair for the suspect; foils' voices are too similar to the suspect  $\rightarrow$  the task is too difficult and unfair for the earwitness

In both 1 & 2 the notion of VOICE SIMILARITY is key: the smaller the distance between voices, the closer they are perceived to be

....but how to measure voice similarity/distances?

#### **OBJECTIVES**

✓ Investigate one under-researched aspect: voice quality (VQ) similarity

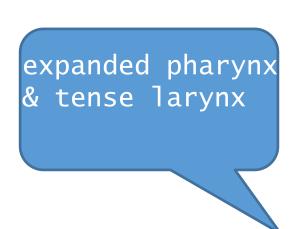


some proposals exist on how to measure speaker similarity for voice parades (de Jong et al. 2015), but investigations focusing on VQ are rare



VQ = quasi-permanent quality of a speaker's voice resulting from a combination of long-term laryngeal and supralaryngeal settings (Laver 1980)





trained expert

Dietmar Wunder is the German voice

of Daniel Craig. Check their similarity

non-expert listener

**MATERIALS & METHODS** 

# Speaker corpus: DyViS

- 99 male speakers (aged 18-25) Standard Southern British English
- semi-scripted telephone conversation
- 7 min. net speech

**VQ protocol:** Vocal Profile Analysis (VPA)

- componential approach to the perceptual assessment of VQ
- VQ = emerging from several components or settings (defined in relation to a 'neutral setting')
- 3 point scale: slight > marked> extreme

#### Perceptual assessment procedure

- 3 trained phoneticians (ESS, PFo, PFr)
- 2-stage methodology: (1) pilot assessment of 10 random subject; (2) calibration meeting
- 99 VPA ratings each (blind procedure)
- cross-coder calibration process

#### **INTERRATER AGREEMENT**

82.6% absolute agreement 89.1% within 1 scalar degree Unweigthed Fleiss' kappa = moderate to substantial

(See San Segundo et al. 2017)

#### **Cluster analyses**

- measured distances: squared Euclidean distances
- 1) hierarchical method (Ward's method)
- because non-hierarchical methods require specifying # clusters
- # clusters determined at the step where distance coeff. show a great diff. in agglomeration schedule  $\rightarrow$  checked visually in the scree diagram (here: two clusters)
- (2) non-hierarchical method (k-means)
- used to actually form the clusters
- # clusters is fixed (two) and an initial set of k 'seeds' (aggregation centers) is provided
- given a certain threshold, all units are assigned to the nearest cluster seed + new seeds are computed until no reclassification is necessary

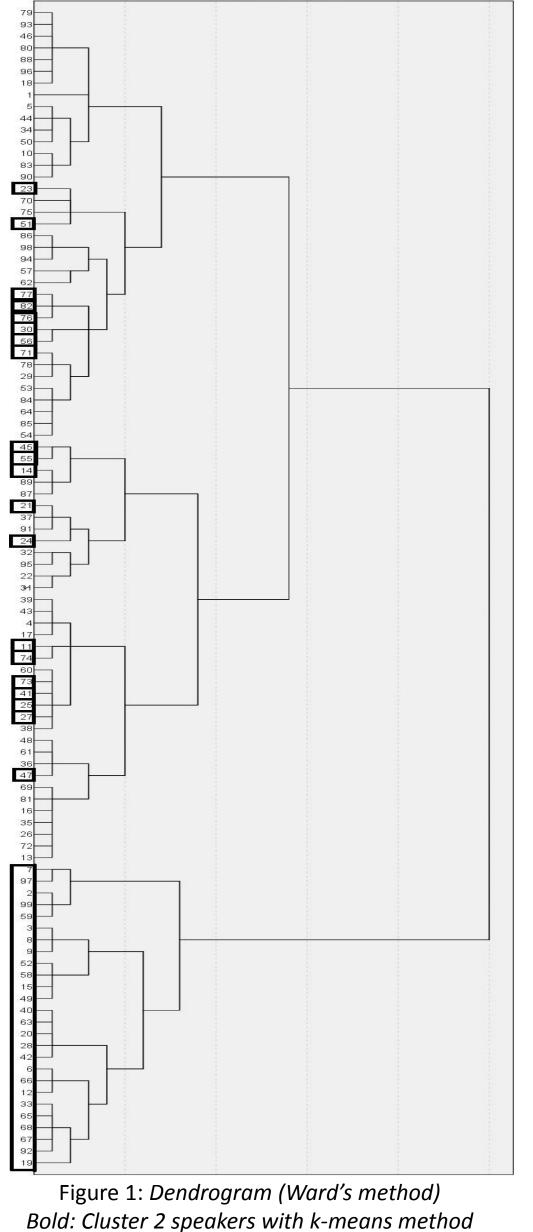
#### **RESULTS**

#### (1) Hierarchical Ward's method

- (-) affected by the order of variables (anatomical progression of VPA settings from lips to larynx)
- (+) allow dendrogram representation / tree structure

#### Non-hierarchical k-means method

- Cluster 1 = 53 speakers; Cluster 2 = 46 speakers
- VPA settings assigned to either Cluster 1 or 2
- settings that contribute most to the separation of the clusters because high average % in one cluster vs. the other (ANOVA test)



Cluster 1 Cluster 2 1. Lip rounding 2. Lip spreading 3. Labiodentalisation 4. Extensive labial range 5. Minimised labial range 6. Close jaw 7. Open jaw 8. Extensive mandibular range .00 9. Minimised mandibular range .04 10. Advanced tongue tip .70 11. Retracted tongue tip .06 1.26 12. Fronted/raised tongue body 1.28 13. Backed/lowered tongue body .00 14. Ext. lingual range .04 15. Min. lingual range 16. Pharyngeal constriction 17. Pharyngeal expansion 18. Nasal 1.08 1.41 .09 19. Denasal 20. Raised larynx .04 1.09 21. Lowered larynx .62 22. Tense vocal tract 23. Lax vocal tract .64 .09 24. Tense Larynx 1.06 25. Lax Larynx 26. Falsetto 1.53 27. Creaky 28. Whispery 1.49 29. Breathy .08 30. Murmur 31. Harsh .70 .11 .00 32. Tremor Table 1: Final cluster centers ANOVA test significance level:

\* p<0.01 \*\* p<0.001

#### **DISCUSSION & CONCLUSIONS**

- Cluster analysis = useful to find patterns in large data sets
- different techniques → slightly diff. results
- hierarchical vs. non-hierarchical → diff. classified speakers

## K-means technique = better method

- ✓ not influenced by order of variables (robust) method even if some cases are dropped)
- ✓ meet requirements for robust classification (Eppler & Stoyko 2011):
  - simple & clear

**CLUSTER** 

Creaky

- meaningful groupings
- consistent with established theories (e.g. Laver 1980: creaky & f0 below 100 Hz)

Breathy

Harsh

- ➤ Clustering methods allow to distinguish perceptually similar speakers (basis: VQ ratings)
- ✓ It was possible to distinguish at least 2 clusters even within a homogeneous population of same-sociolect speakers
- ✓ Forensic implication → annotate databases (VQ info) prior to voice parade
- automatized search of most similar foils to suspect - optimize resources by law enforcement agencies (time and money save)

- minimize subjectivity involved in voice selection

## ➤ Novelty: use of perceptual ratings

 Previous studies: acoustic measures (disadvantage: not necessarily perceptually salient for listeners in judging similarity)

#### > Future:

✓ To which extent auditory expert ratings are comparable with similarity ratings by naive listeners (San Segundo et al. 2016)? Whisper

> Testing other data reduction techniques for data mining (e.g. factor analyses)

# REFERENCES

**CLUSTER** 

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#### **ACKNOWLEDGEMENTS & CONTACT**

This research was funded via the UK AHRC grant Voice and Identity (AH/M003396/1)

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