



Introduction

- Most neuroimaging studies of speech production look at the average neural activity across a group of subjects during a particular speaking task compared to baseline.
- When studying speech disorders that result from idiosyncratic lesions such as stroke or traumatic brain injury, it may be more appropriate to focus on *individual activity patterns*.
- As a first step toward characterizing the speech production network in individuals post-stroke, we sought to quantify the reliability of speech activation in healthy individuals across multiple sessions and speech tasks.
- Brain activity from healthy subjects who participated in two similar speech production studies in our lab was compared to assess reliability across scanning sessions.

Methods

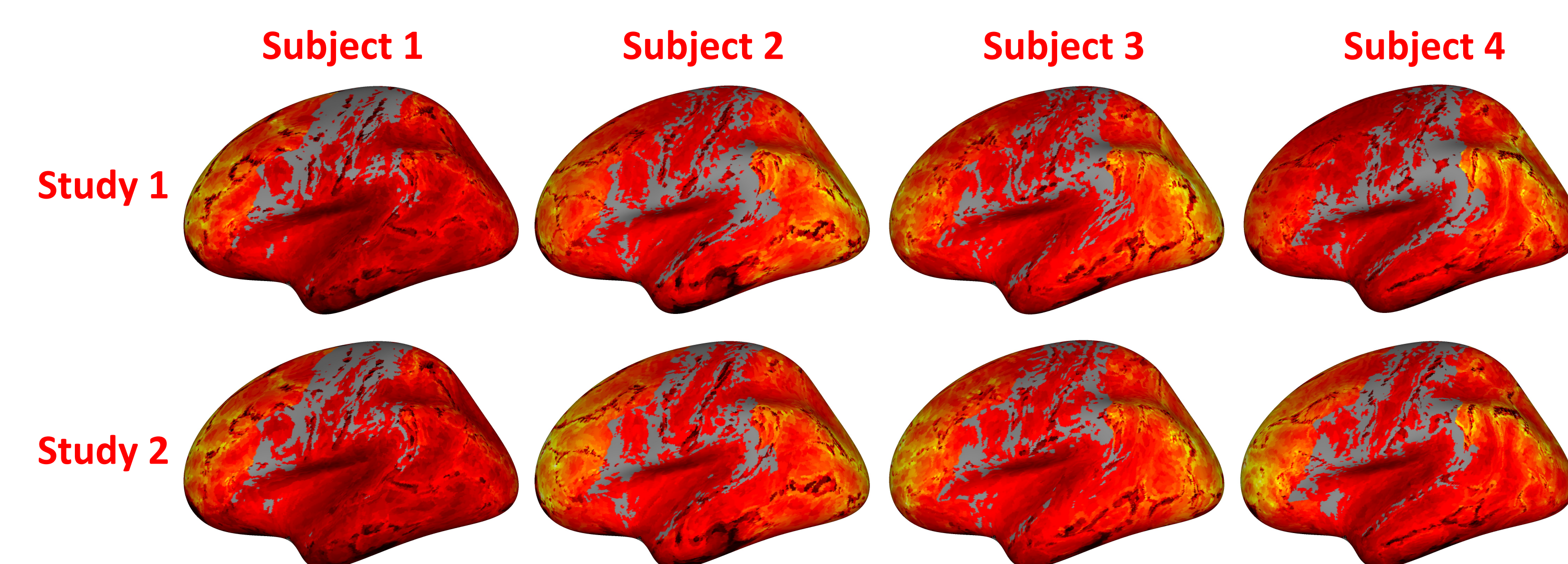
Data

- Test Set:** 14 subjects (7M/7F) participated in two pseudoword production fMRI studies; 112 *speech* > *baseline* contrasts, 28 mean functional masks
 - Mean speaker age: 29 (19-44)
 - Days between sessions: 13 (6-52)
 - Effect-size maps were **masked to only include areas in the speech network**, and thresholded at **the highest 10% of vertices**
- Distractor Set:** 62 subjects (119 *speech* > *baseline* contrasts, 68 mean functional maps) were processed similarly to be used for training and testing a classifier.

Analyses

- Classifier analyses (see figure 5)
- Similarity measures:
 - Dice coefficient (Bennett and Miller, 2010)
 - Whole-brain intraclass correlation coefficients calculated using ICC(C, 1) from McGraw and Wong (1996).

Fig. 1: Baseline: Mean Functional Activation

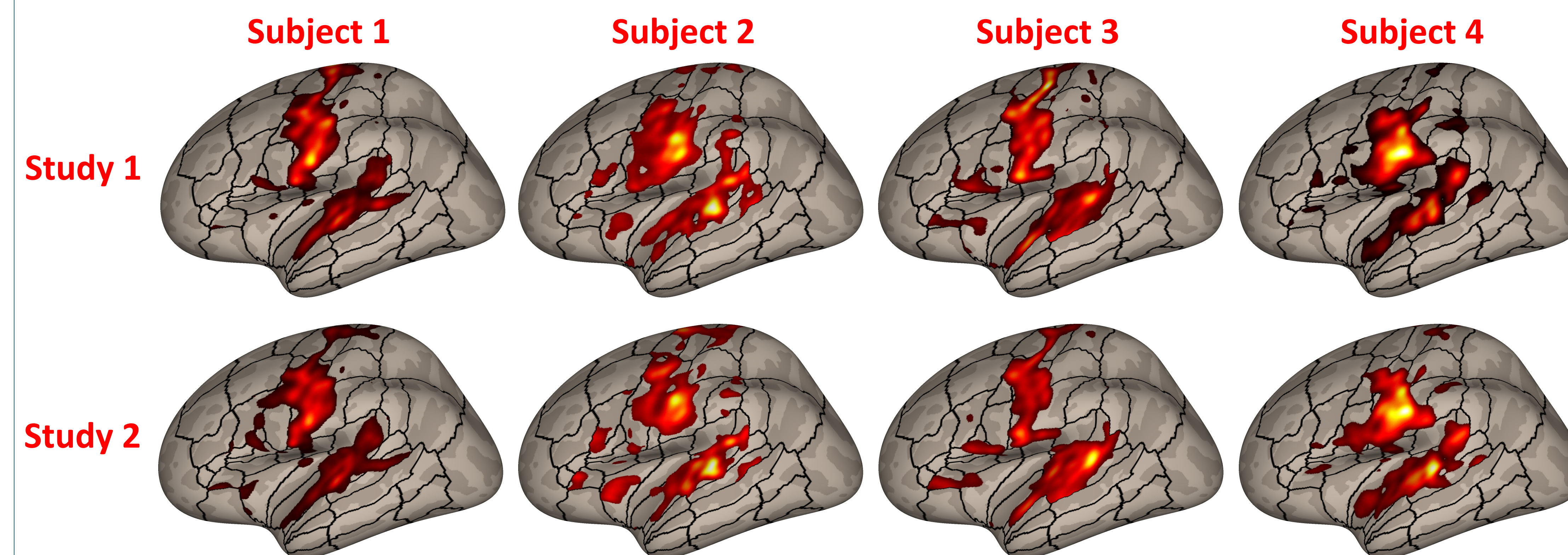


- Mean BOLD signals (thresholded data in grey used for analysis).
- We classified test subjects with **100% accuracy** from among **76 subjects**, using **25 principal components**.
- Subjects' average functional BOLD signals were **highly reliable**: ICC and **Dice coefficient** were both **0.71**.
- **The ICC across subjects was 0.57**.

References

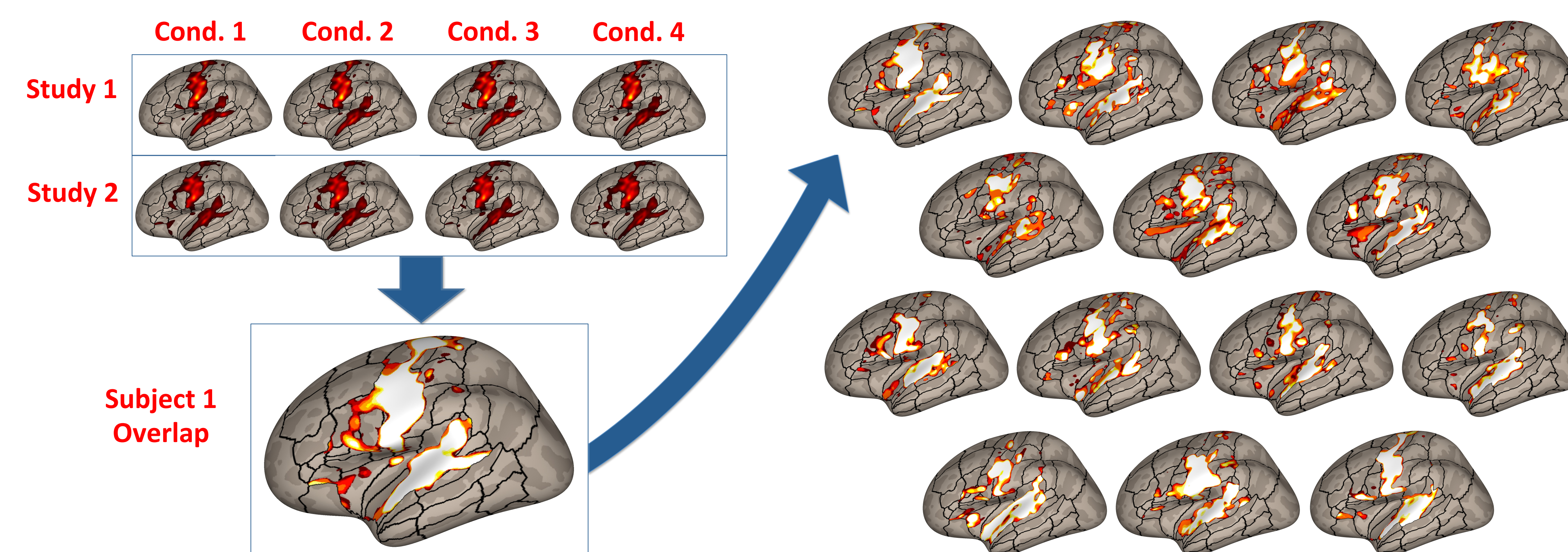
- Bennett, C.M., & Miller, M.B. (2010). How reliable are the results from functional magnetic resonance imaging? *Annals of the New York Academy of Sciences*, 1191(1), 133-155.
- McGraw, K., & Wong, S (1996). Forming inferences about some intraclass correlation coefficients. *Psychological Methods*, 1(1), 30-46.

Fig. 2: Speech-Baseline Activation



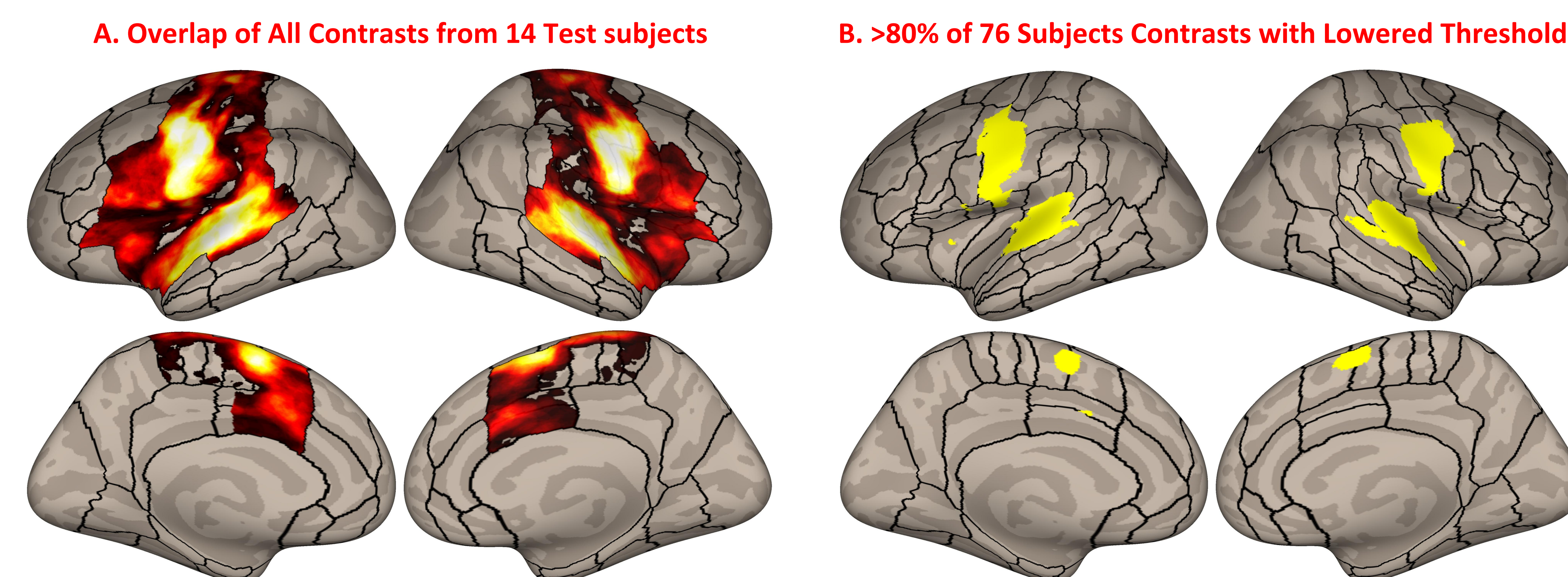
- Speech activation is **highly reliable** for each subject across two studies.
- We could classify individuals with **100% accuracy** from among **76 subjects**, using only **25 principal components**.

Fig 3. Single Subject Overlap



- For speech activation, average **intra-subject ICC** was **0.88** as compared to an **inter-subject ICC** of **0.64**. Average intra-subject **Dice coefficient** was **0.73**.
- As the right side shows, each subject has a unique pattern of intra subject overlap.

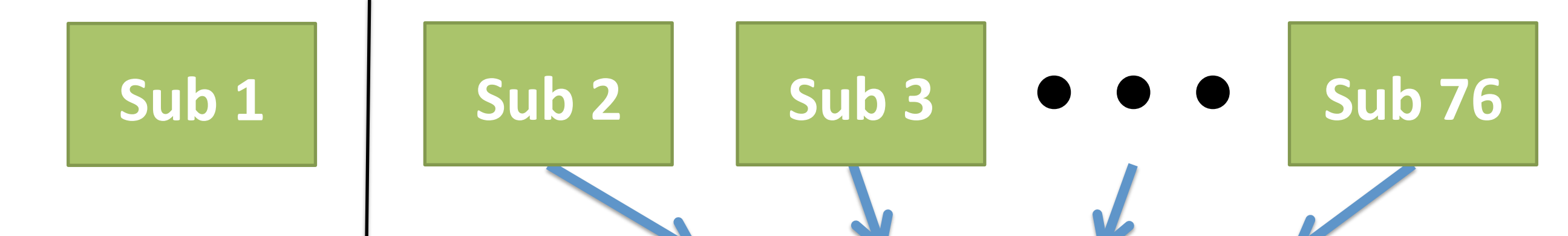
Fig. 4: Grand Overlap Speech Maps



- Most subjects showed large areas of activation in ventral motor and premotor cortex, auditory cortex, and medial premotor cortex during speech production.

Fig 5. Classifier Analysis

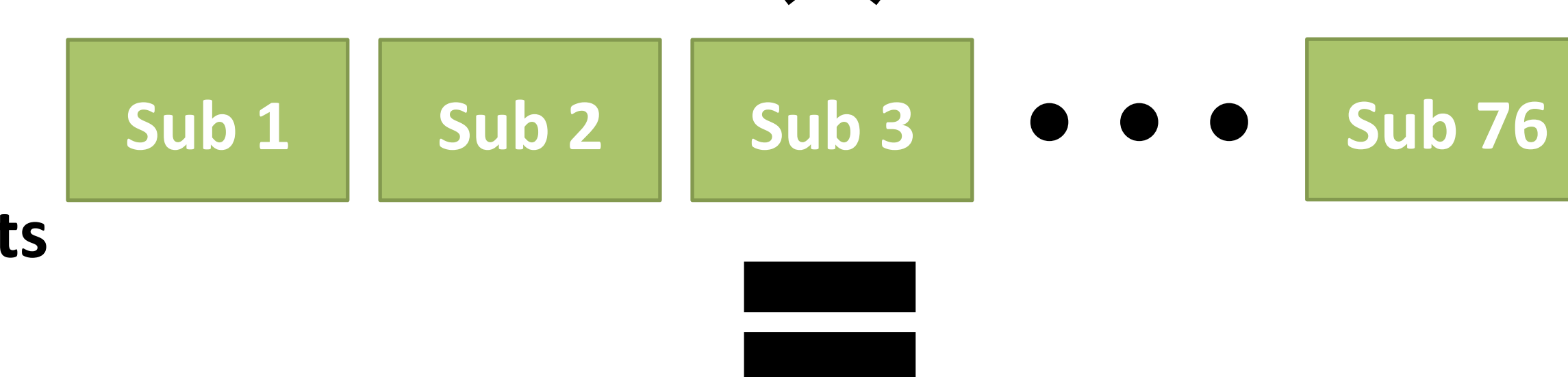
1. Remove one 'test subject'



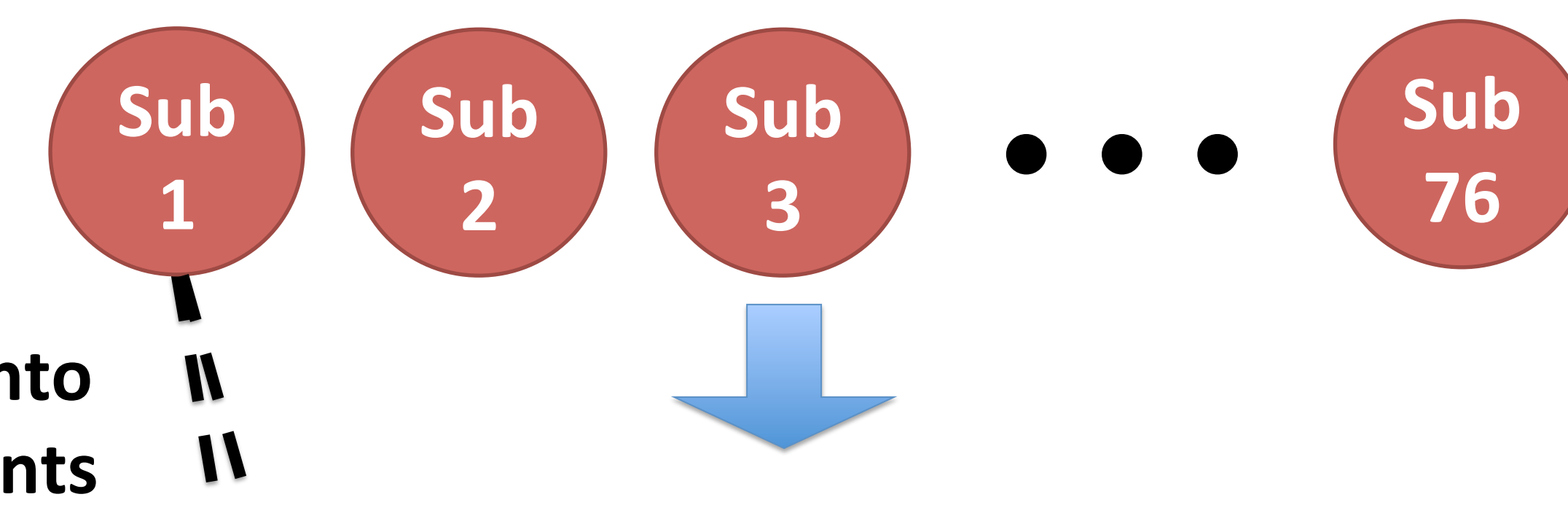
2. Perform PCA on the rest



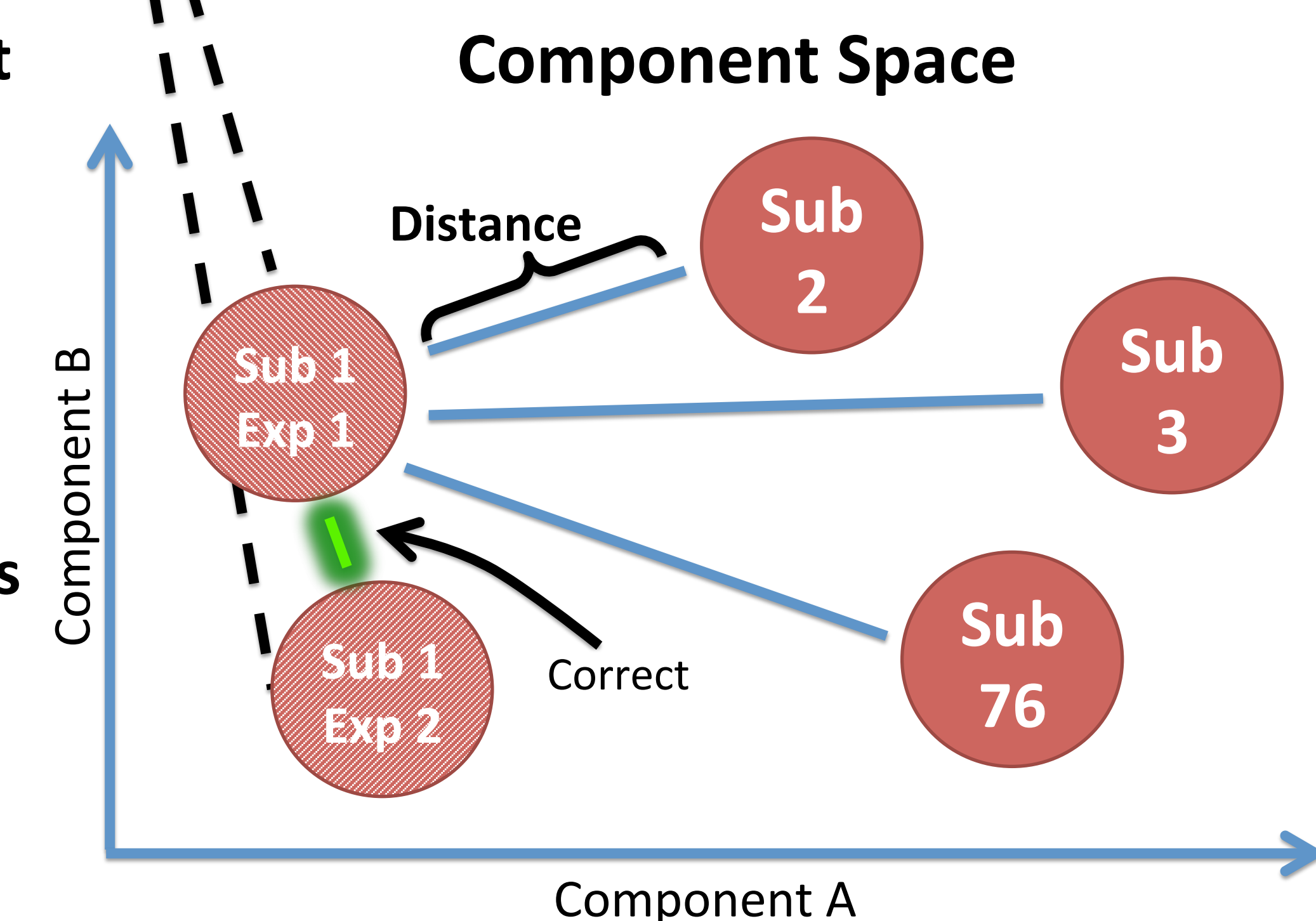
3. Apply principal components to all subjects



4. Split test subject data into two experiments



5. Compare test subject in Exp 1 and all subjects (using Exp 2 for same subject)



6. Test subject is classified as the closest subject

7. Repeat 1-6 for each test subject

Conclusions

- Classifier success suggests that intra-subject variability is smaller than inter-subject variability.
- Subjects' mean functional and speech > baseline activation are highly reliable across scanning sessions
- This suggests that healthy speakers have a unique neural "fingerprint" that can be observed during speech production.
- Over 80% of contrasts from 76 subjects showed activity in large swaths of common speech production areas.
- Overall, this suggests that single-subject studies are likely to yield reliable results.

Acknowledgements

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