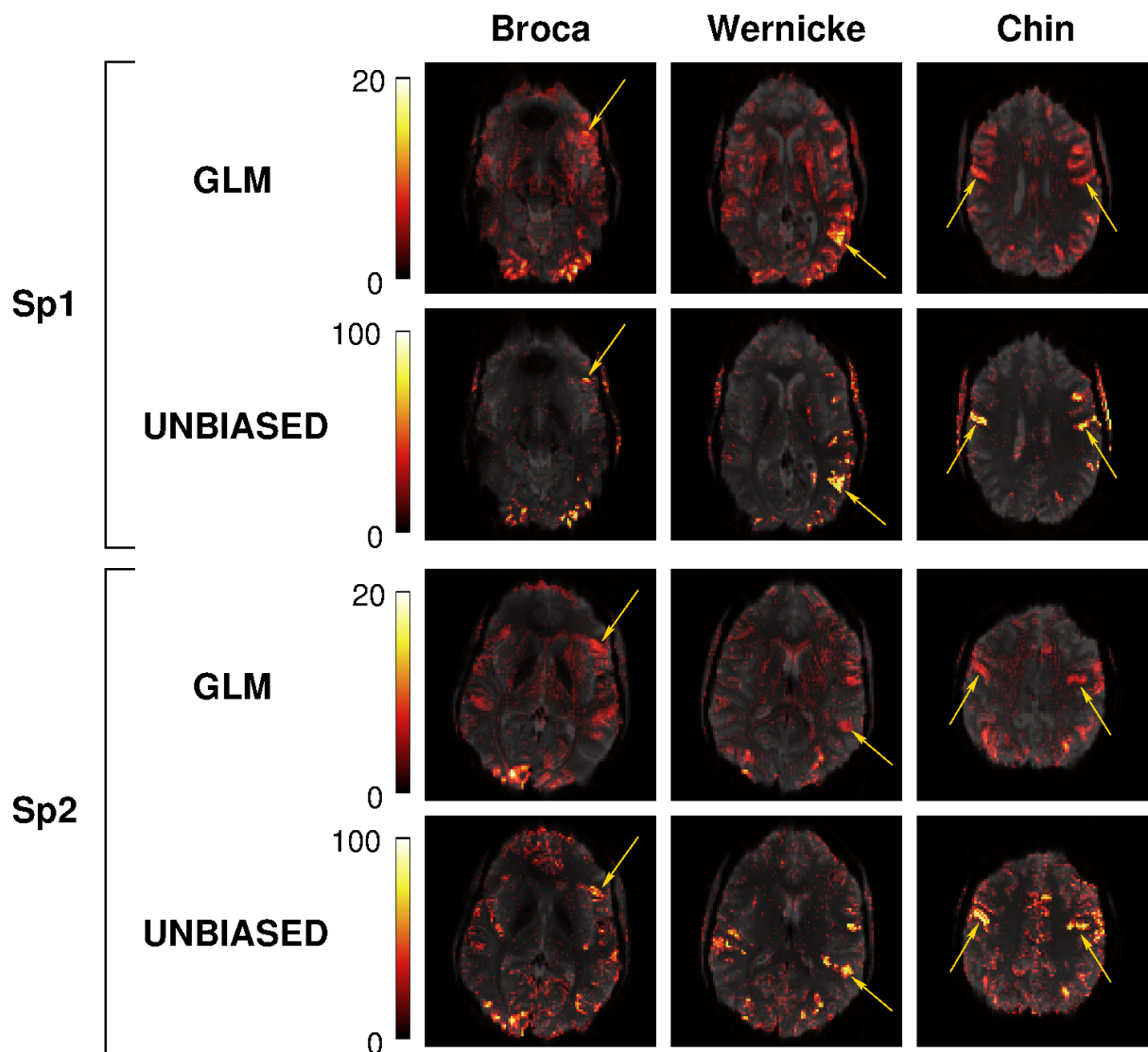
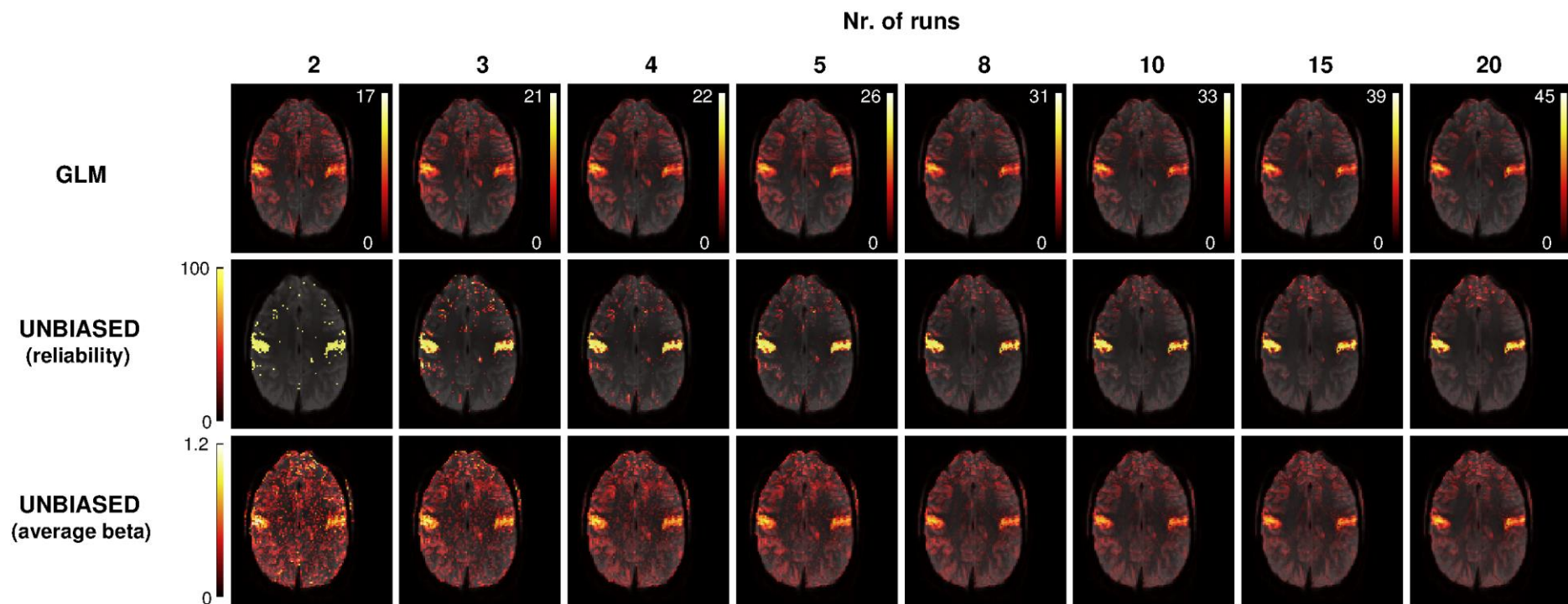


SUPPORTING MATERIAL



**Supporting Fig. 1** GLM (*1<sup>st</sup>* and *3<sup>rd</sup>* rows) and UNBIASED (*2<sup>nd</sup>* and *4<sup>th</sup>* rows) results for two healthy volunteers performing an overt speech task in an ABABABA block design with the same timing as the motor tasks in this study. Slices containing functional areas responsible for speech processing (*1<sup>st</sup>* and *2<sup>nd</sup>* columns) and chin motion (*3<sup>rd</sup>* column) are shown. Results indicate that the method is also effective in cognitive functional regions, which produce smaller and less reliable signal changes than motor or visual areas (e.g. Broca and Wernicke areas responsible for speech processing).



**Supporting Fig. 2** Effect of increasing the number of runs in the GLM (*top row*) and UNBIASED (*middle row*) analyses of data from a healthy volunteer performing a chin task. *Bottom row*: UNBIASED average beta (fit) values. Results indicate that UNBIASED is able to accurately depict activation with a number of runs as low as two (the minimum required for inter-run fitting of voxel time courses). The gradation (or number of intervals) of reliability values increases with the number of runs (e.g. there are only two reliability levels (0 and 100) if two runs are used). If the number of runs is low, average beta (fit) values could be used to aid the user in distinguishing true activation from artefacts.