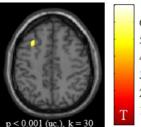
## Wissenschaftlicher Bericht





$$x = -30, y = 20, z = 46$$

Left medial frontal cortex

T = 6.32, MNI: x = -30, y = 20, z = 46

Voxel-level:  $p_{uncorr} < 0.001, p_{FWE-corr} = 0.991$ Cluster-level:  $p_{uncorr} = 0.003, p_{FWE-corr} = 0.062$ 

Figure 4. Exemplary application of the RFT-based power, PPV, and sample size calculation framework. (A) The upper panel depicts the results of a perceptual decision-making pilot study with n = 10 participants for contrasting perceptual choices based on high and low visual sensory evidence. The T-values from the identified cluster in the left medial frontal gyrus were averaged to obtain a raw effect size testimate, which was then adjusted based on the effect size bias estimates reported in Figure 7 of Genter et al. (2018) and reproduced in the lower subpanel of panel (A). (B) Sample size calculations for voxel-level minimal and maximal power and PPV based on the effect size estimates of the pilot fMRI study. (C) Sample size calculations for cluster-level minimal and maximal power and PPV based on the effect size estimates of the pilot fMRI study. (C) Sample size calculations for cluster-level minimal and passimal power and PPV based on the effect size estimates of the pilot fMRI study. (S) rimplementational details, please see right, figure 4.m.

revealed a cluster of activity in the left medial frontal gyrus, as shown in the upper panel of Figure 4A (for further details about the experimental and data-analytical procedures, please see Supplement S.5). Our aim was to use the effect size estimate derived from this cluster to calculate the sample sizes necessary to achieve minimal and maximal power and PPV levels of 0.8 for corrected voxel- and cluster-level inference at a significance level of  $\alpha'_{\rm PWE}=0.05$ , a partial alternative hypothesis parameter of A=0.1, and a prior hypothesis parameter of A=0.2. To this end, we evaluated the average T-values of the cluster, yielding T=4.65, which translates into an effect size estimate of  $\hat{d}=4.65/\sqrt{10}=1.47$ . However, it is well known that effect size estimates resulting from the thresholding of mass-univariate statistical parametric maps exhibit biases (e.g., Vul et al., 2009; Poldrack et al., 2017). To correct our effect size estimate for this bias, we capitalized on recent results by Geuter et al. (2018), which are depicted in the lower panel of Figure 4A. Specifically, using task-related fMRI data from the Human Connectome Project 500 (Van Essen et al., 2013), Geuter et al. (2018) estimated the effect size bias exhibited by activations detected in random data subsets of 10 to 100 participants from the approximately 500 participants. As reported in Figure 7A of Geuter et al. (2018) and visualized in the lower





## Reproduktion

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```
% This function implements the sampling of Gaussian random fields with
% Gaussian covariance functions as visualized in Figure 2 of Ostwald et al.
% (2018) On parametric p-values.
   Author - Dirk Ostwald
% initialization
close all
% random number generator setting for reproducible results
 addpath(genpath(fullfile(pwd, "Utilities")))
                                                                   ; % RFP utilities
 addpath(genpath(fullfile(pwd, SPM))))
                                                                   ; % defaults
figures
fdir = fullfile(fileparts(pwd), [Figures])
                                                                   ; % figure director;
% discrete domain D = [x_1_1,..,x_1_n] \times [x_2_1,...,x_2_n]
                                                                   ; % number of discrete realizations per dimension
     = K^2
                                                                   ; % number of discrete realizations
                                                                   ; % domain minimum
D_min = 0
                                                                   ; % domain maximum
D_max = 1
       = linspace(D_min,D_max,K)
                                                                   : % x ordinates
       = linspace(D min,D max,K)
                                                                   ; % y ordinates
                                                                   ; % discrete domain initialization
         sub-005 con 0001.nii
                                                                                  2019-03-26 01:51 PM
         sub-006_con_0001.nii
                                                                                  2019-03-26 01:51 PM
         sub-007_con_0001.nii
                                                                                  2019-03-26 01:51 PM

    sub-008 con 0001.nii

                                                                                  2019-03-26 01:51 PM
         sub-009_con_0001.nii
                                                                                  2019-03-26 01:51 PM

    sub-010 con 0001.nii

                                                                                  2019-03-26 01:51 PM
```