

1 PROBLEM SETTING

- Simultaneously testing thousands of voxels for activation leads to an inflation of false positives (**the multiple testing problem**).
- Several (voxelwise) corrections to protect the amount of false positives exist, for example:
 - o Bonferroni (BF): protects the family-wise error rate
 - o Benjamini-Hochberg (BH): protects the false discovery rate
- Multiple testing procedures are evaluated on their average performance with respect to error rates.
- **What about the variability on the results?**
- The higher the **selection variability**, the lower the **selection stability**.
- Qiu et al. (2006) show that multiple testing procedures can be highly unstable.

2 GOAL

1. Using bootstrap procedures, we **measure selection variability** on test results for multiple testing corrections following BF (FWER) and BH (FDR).
2. We present a **new testing strategy** which includes both **significance** and **selection variability** in the decision criterion (Gordon et al., 2009).
3. The new procedure is evaluated through Monte Carlo simulated fMRI images.

Does the new procedure improve the selection stability?

3 METHODS

Creation of images

We create 2-dimensional images consisting of 64x64 voxels. Activation is placed in the center of the image (32x32 voxels). We consider a 20s ON/ 20s OFF block design (TR=1) repeated 3 times and convolved with a canonical HRF. Temporal and spatial noise is added.

GLM

Before voxelwise linear regression of the measured signal on the signal components, data are pre-whitened to account for the temporal correlation using the procedure described by Worsley et al. (2003).

voxel thresholding

p-values are thresholded according to BF and BH

New testing strategy and voxel selection

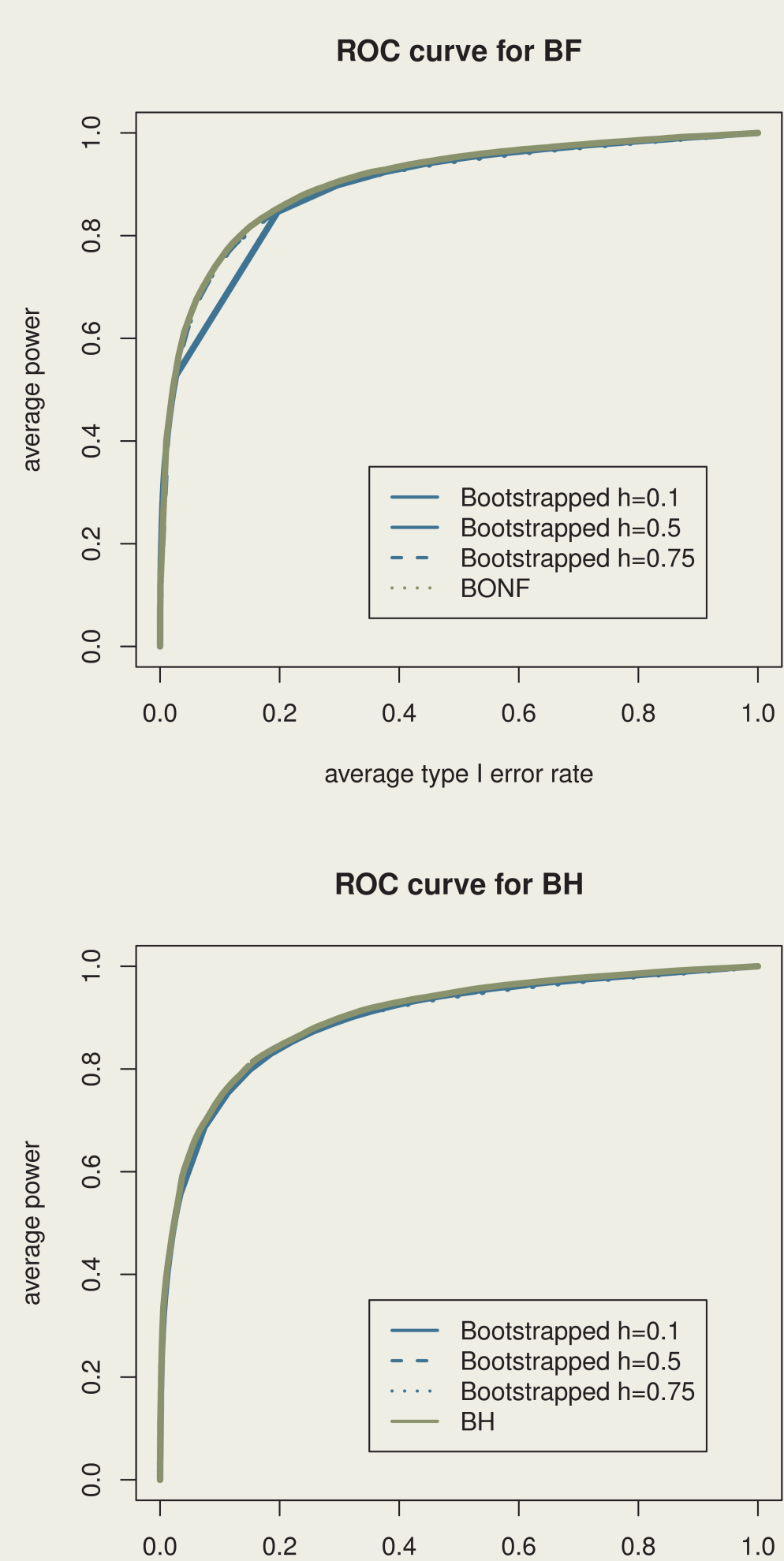
- 100 bootstrap samples of size $n=120$ are taken from the residuals of the voxels.
- The bootstraps are used to construct 'replicates' for the time series of images, by re-adding the temporal correlation.
- Each replicate obtained by bootstrapping is analysed as the original time series.
- 4 different criteria for voxel selection:
 - (1) the original multiple testing procedure (MTP) and selected in at least
 - (2) $h = 10\%$, (3) $h = 50\%$, (4) $h = 75\%$ of the replicates using the MTP.

simulation

- We create 500 time series of images
- The average performance over 500 simulations is investigated

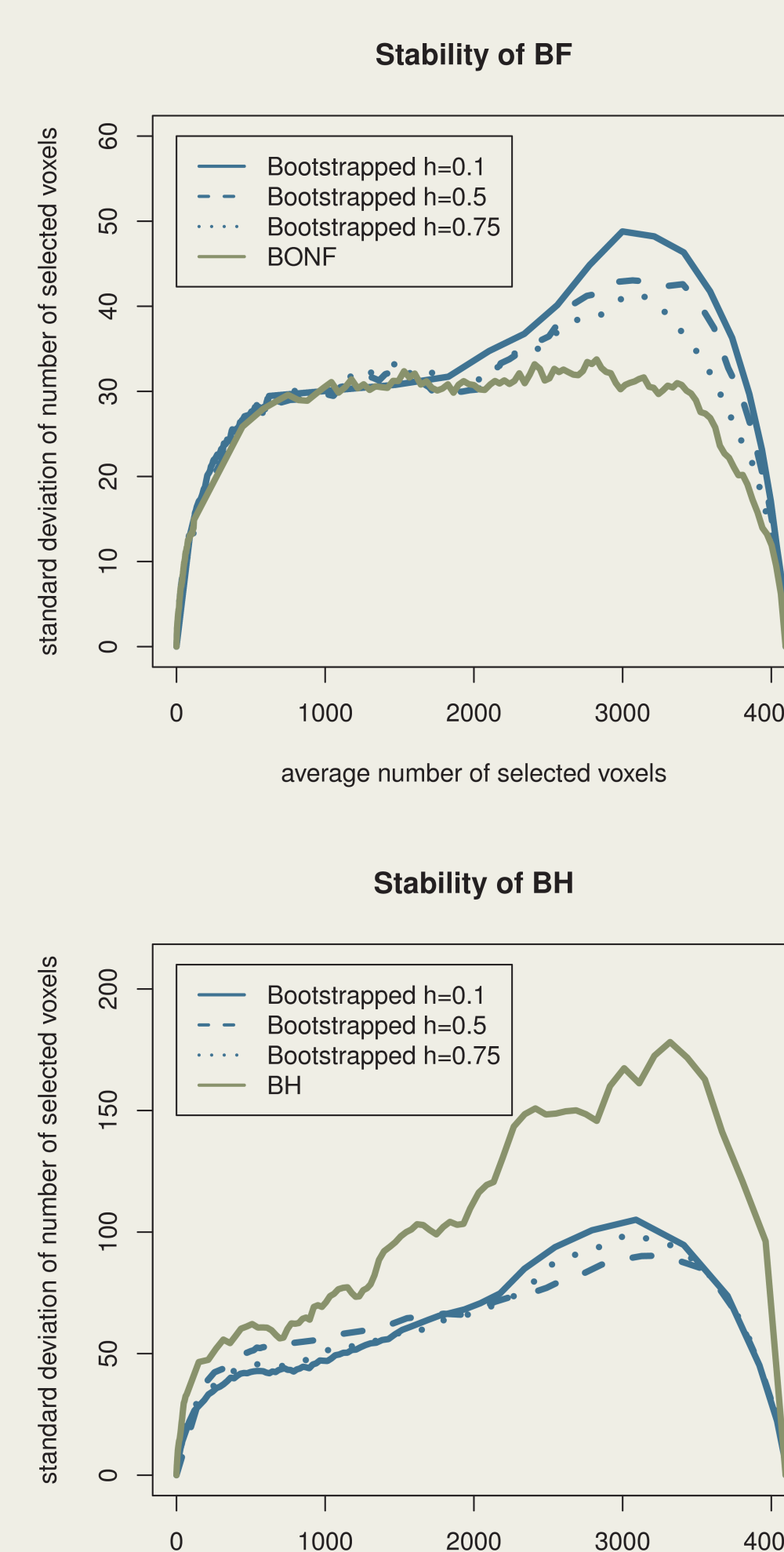
4 RESULTS

ROC-CURVES



For BF, BH and their bootstrap counterparts, the relation between sensitivity and specificity is the same.

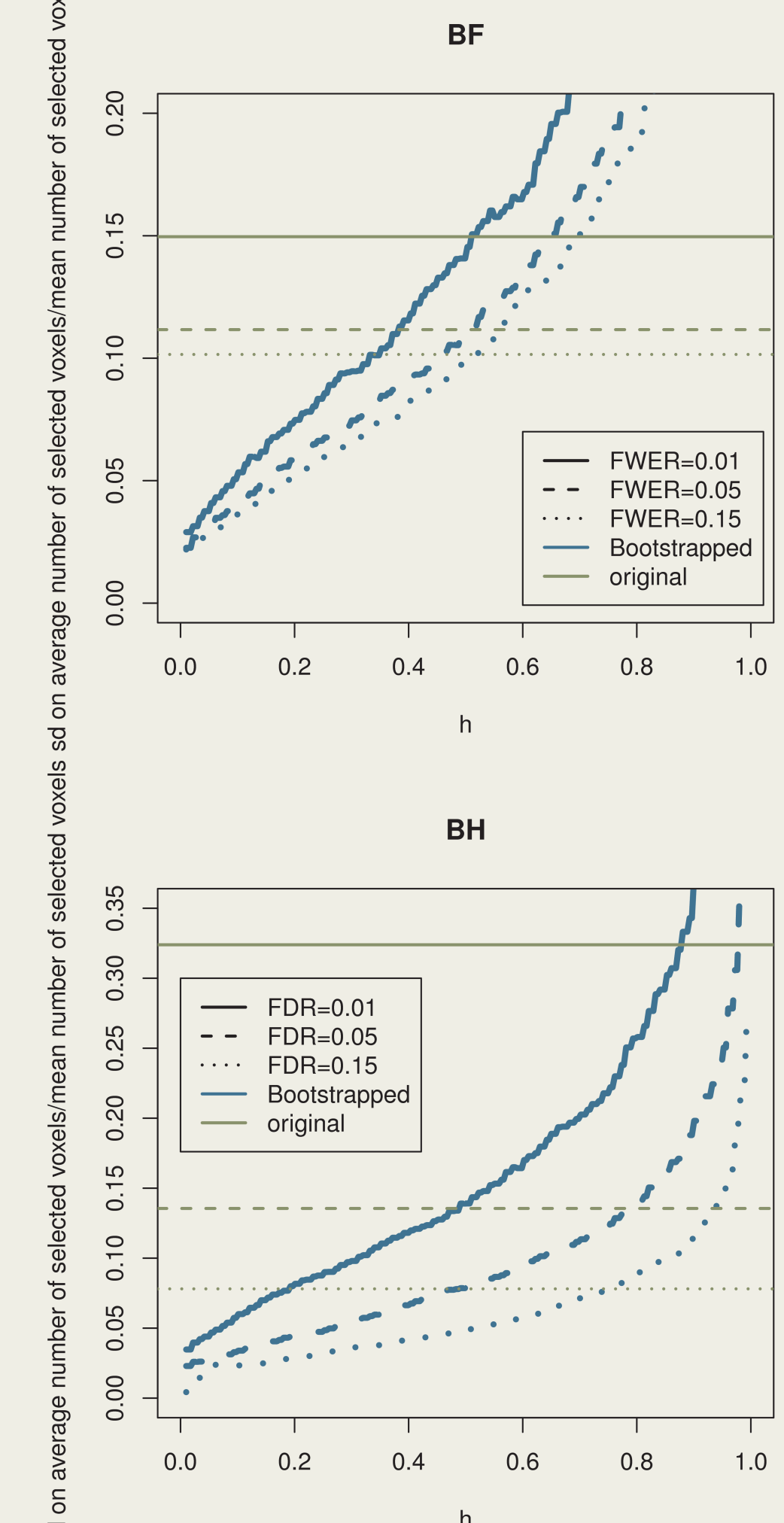
SELECTION VARIABILITY



BF: Variability on test results is not reduced by including selection variability into the decision criterion.
BH: - less stable than BF
 - the stability on the test results can be strongly improved using the bootstrap procedure, up until the level of stability of BF

THE ROLE OF H

h = threshold for selection frequency in bootstrap samples



The more stringent the selection criterion, the higher the variability on test results.

5 CONCLUSION

- We presented a bootstrap approach that can be used with any multiple testing procedure to improve selection stability.
- For BF and BH, the trade-off between power and specificity is maintained, while the stability for BH is improved
- The choice of an optimal h is a topic for further research

REFERENCES

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- Qiu, X., Xiao, Y., Gordon, A. & Yakovlev, A. (2006). Assessing stability of gene selection in microarray data analysis. *BMC Bioinformatics* 7(50).
- Worsley, K. J. (2003). *Statistical analysis of activation images*, Oxford university press, chapter 14, pp 251-270.

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