

# Appendix 1. Stability selection

## Methods

We applied stability selection (Meinshausen and Bühlmann 2010, Shah and Samworth 2013) to identify base-learners, and thus covariates, that were commonly selected in the majority of 100 random subsamples of the data. We set the number of selected base-learners per boosting model ( $q$ ) to 35 and established an upper bound of two for the per-family error rate (PFER; Meinshausen and Bühlmann 2010, Shah and Samworth 2013, see also Hofner et al. 2015 for details in the context of boosting) which, given the 48 base-learners in the occupancy and count models (see Equation 1 in manuscript), corresponded to an upper bound of  $\alpha = 0.042$  for the per-comparison error rate. **Why did we choose  $q = 35$ ? Describe complementary pairs subsampling and our choice of unimodal vs. r-concavity assumptions for the two model type (occupancy vs. count)? Any other relevant details to include?**

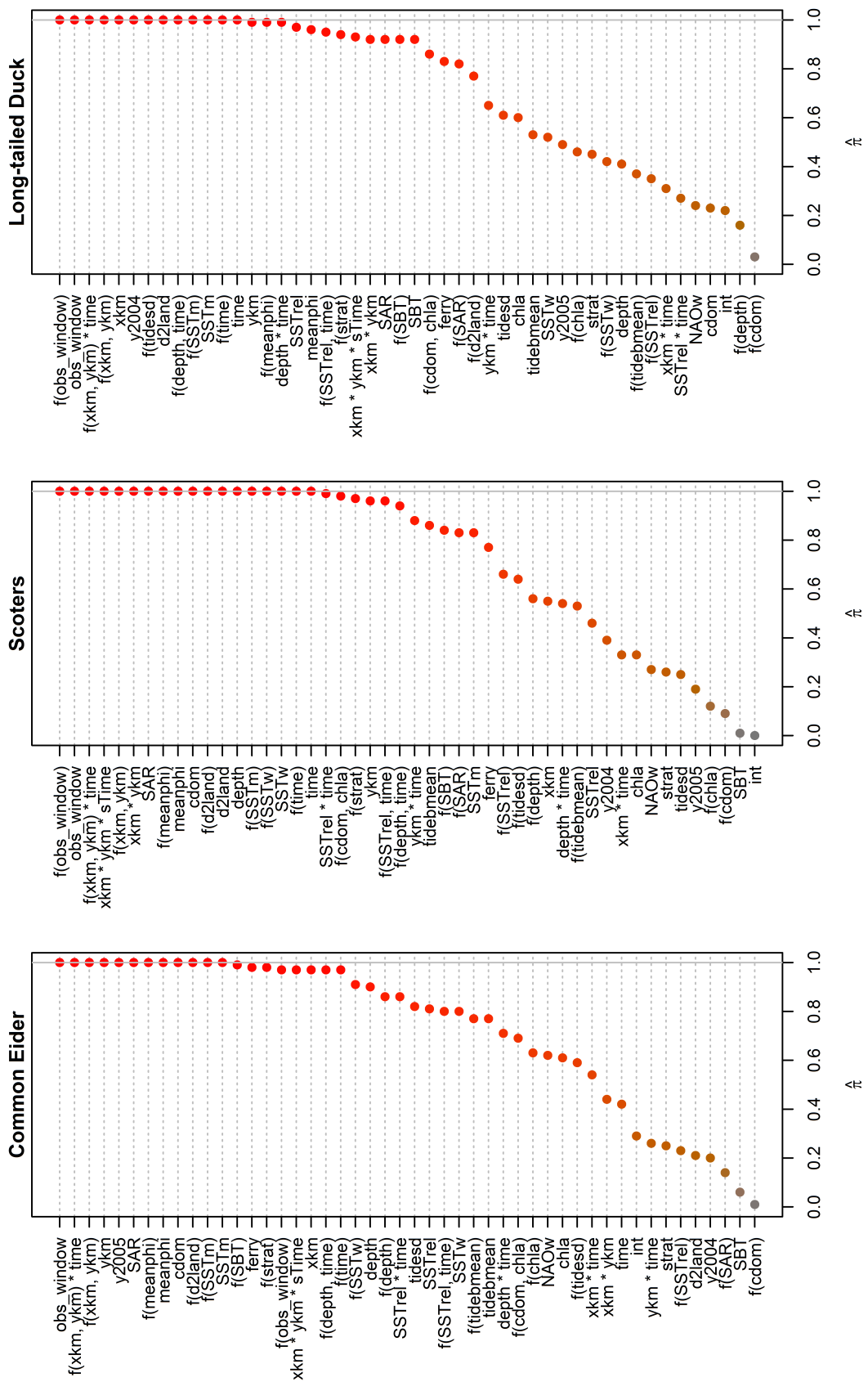
## Results

### Occupancy models

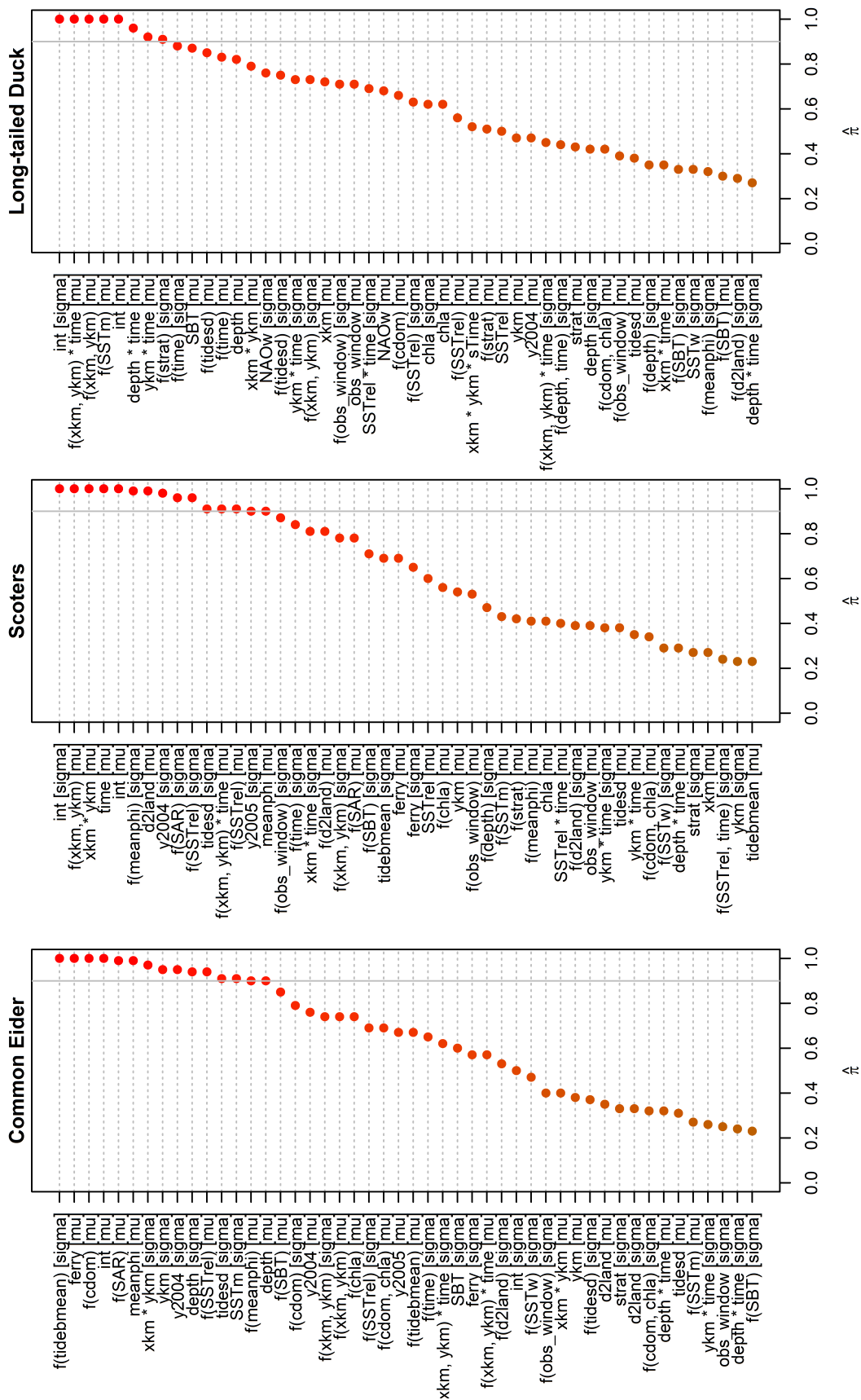
Given our specifications ( $q = 35$ ; PFER upper-bound = 2), only base-learners selected in all 100 subsamples (i.e.,  $\hat{\pi} = 1$ ) were identified as stable (Figure 1.1).

### Count models

Given our specifications ( $q = 35$ ; PFER upper-bound = 2), only base-learners selected in at least 90 of the 100 subsamples (i.e.,  $\hat{\pi} = 0.9$ ) were identified as stable; this threshold applies to the selection of base-learners for the conditional mean ( $\mu$ ) and conditional overdispersion ( $\sigma$ ).



**Figure 1.1** Stability selection using complementary pairs subsampling and unimodality assumption for sea duck occupancy models. The number of selected base-learners in each model run was set to  $q = 35$ . Base-learners with selection frequencies above the threshold ( $\hat{\pi}$ ; vertical gray line) were considered stable with upper bound  $\text{PFER} = 2$ .



**Figure 1.2** Stability selection using complementary pairs subsampling and r-concavity assumption for sea duck conditional count models. The number of selected base-learners in each model run was set to  $q = 35$ . Base-learners with selection frequencies above the threshold ( $\hat{\pi}$ ; vertical gray line) were considered stable with upper bound PFER = 2. Only the top 48 base-learners are illustrated. Brackets indicate the parameter (conditional mean,  $\mu$ , or overdispersion,  $\sigma$ ) to which the base-learner applies.

## Literature cited

- Hofner, B., L. Boccuto, and M. Göker. 2015. Controlling false discoveries in high-dimensional situations: Boosting with stability selection. BMC Bioinformatics in press.
- Meinshausen, N., and P. Bühlmann. 2010. Stability selection (with discussion). Journal of the Royal Statistical Society: Series B (Statistical Methodology) 72:417–473.
- Shah, R. D., and R. J. Samworth. 2013. Variable selection with error control: Another look at stability selection. Journal of the Royal Statistical Society: Series B (Statistical Methodology) 75:55–80.