

# How well do different statistical frameworks predict species- and community-level patterns?

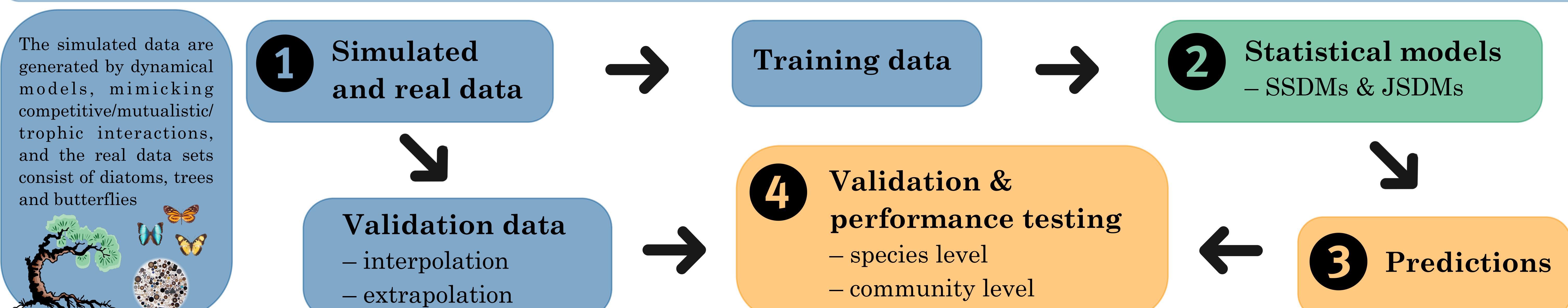


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The idea is to compare various statistical frameworks in terms of their predictive performance



## 2 Models

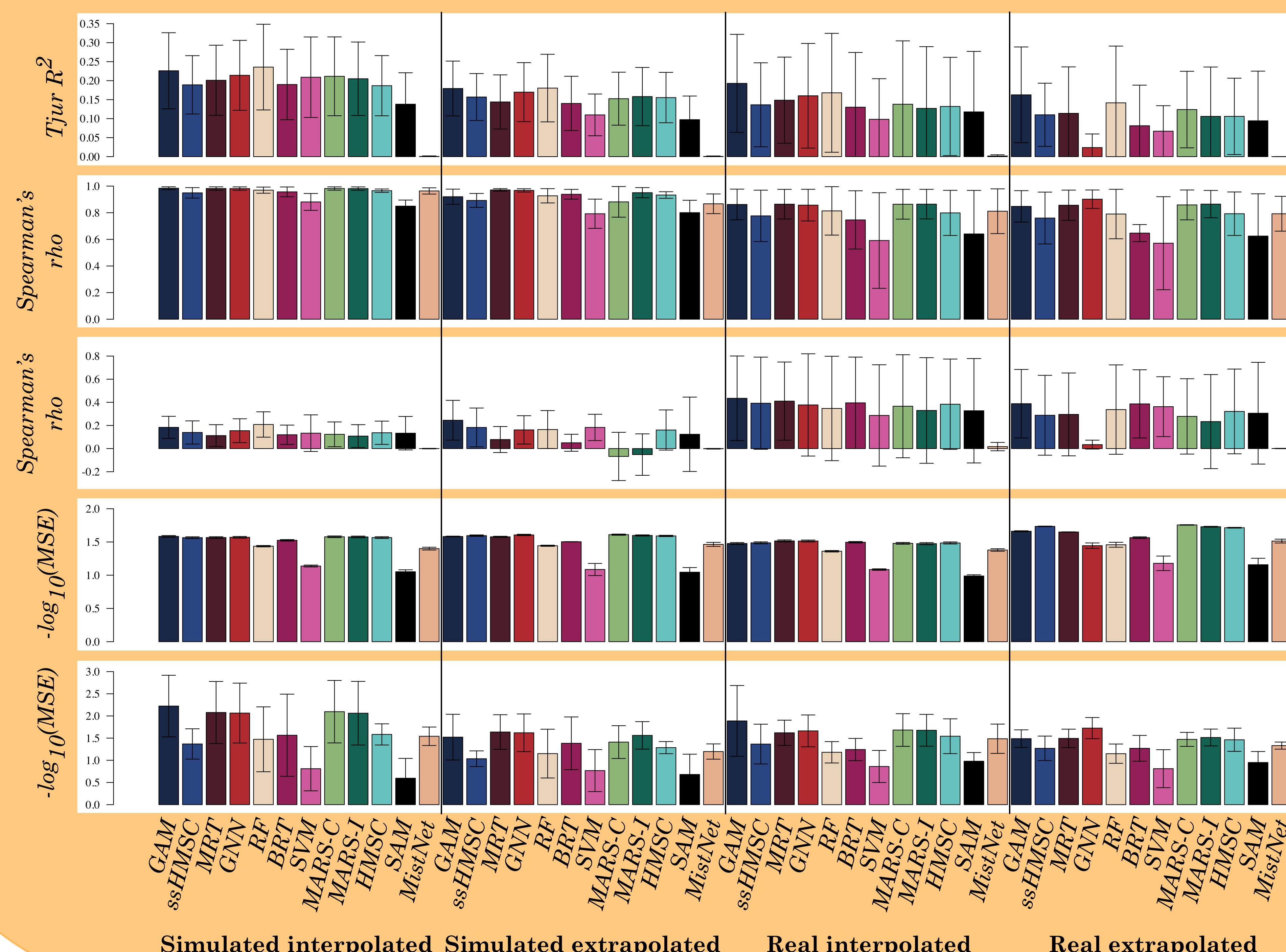
### Stacked species distribution models

- Generalized additive models<sup>1</sup>
- Generalized linear models<sup>2-5</sup>
- Multivariate regression trees<sup>6-7</sup>
- Gradient nearest neighbour<sup>8</sup>
- Random forests<sup>9</sup>
- Boosted regression trees<sup>10</sup>
- Support vector machines<sup>11</sup>
- Multivariate adaptive regression splines<sup>12</sup>

### Joint species distribution models

- Hierarchical modelling of species communities<sup>2-5</sup>
- Species archetype models<sup>13-15</sup>
- Stochastic feedforward neural network<sup>16</sup>

## 3 4 Predictions, validation & performance



We conclude, that even though there are differences in the predictive performance between the models, we so far see no clear division between stacked and joint models in general. Most models seem to perform consistently well, but some perform varyingly regarding accuracy and generality (e.g. GNN, MARS, MistNet). Further development in both model fitting procedures and performance testing is needed in order to obtain some more conclusive results.

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