



Generalizing DBH and height prediction in coast Douglas-fir and red alder

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Forest Inventory: Traditional

- Driven by ground measurement
- Method: Sampling (dominated by StRS)
- DBH: main dendrometric attribute
- Derived attributes:
 - Height
 - Volume



Forest Inventory: Next Generation

- Driven by remote sensing
- Method: population or sophisticated sampling
- Total height: main dendrometric attribute
 - Projected crown area
- Derived attributes:
 - DBH
 - Volume



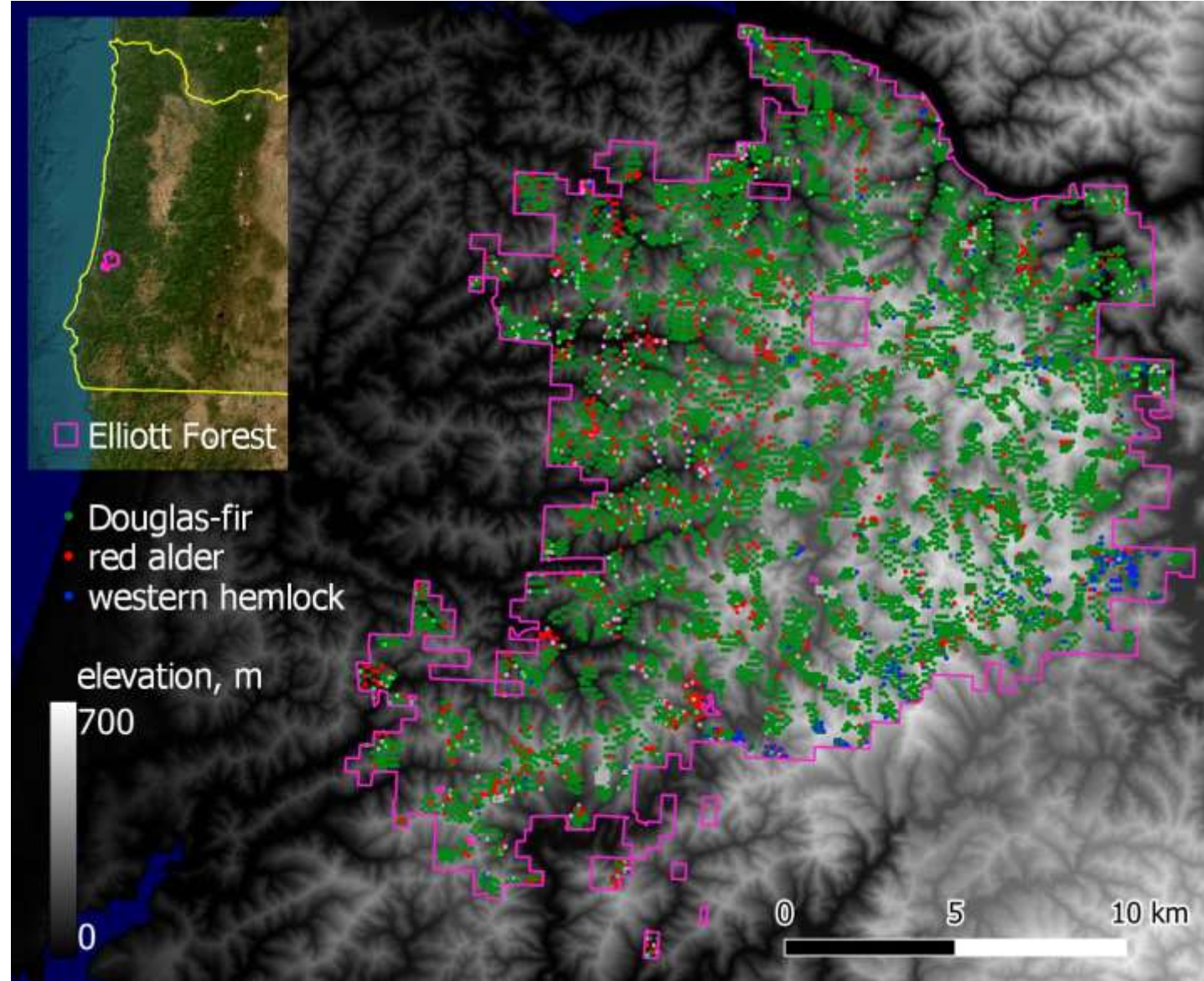
Objective

- Tree-level DBH-Height equations for Douglas fir and red alder
 - Two major Pacific Northwest species
- Focus on:
 - Accuracy and parsimony
 - Coherence: simultaneous and invertible relationships
 - Computational efficiency



Study Area: Elliott State Research Forest

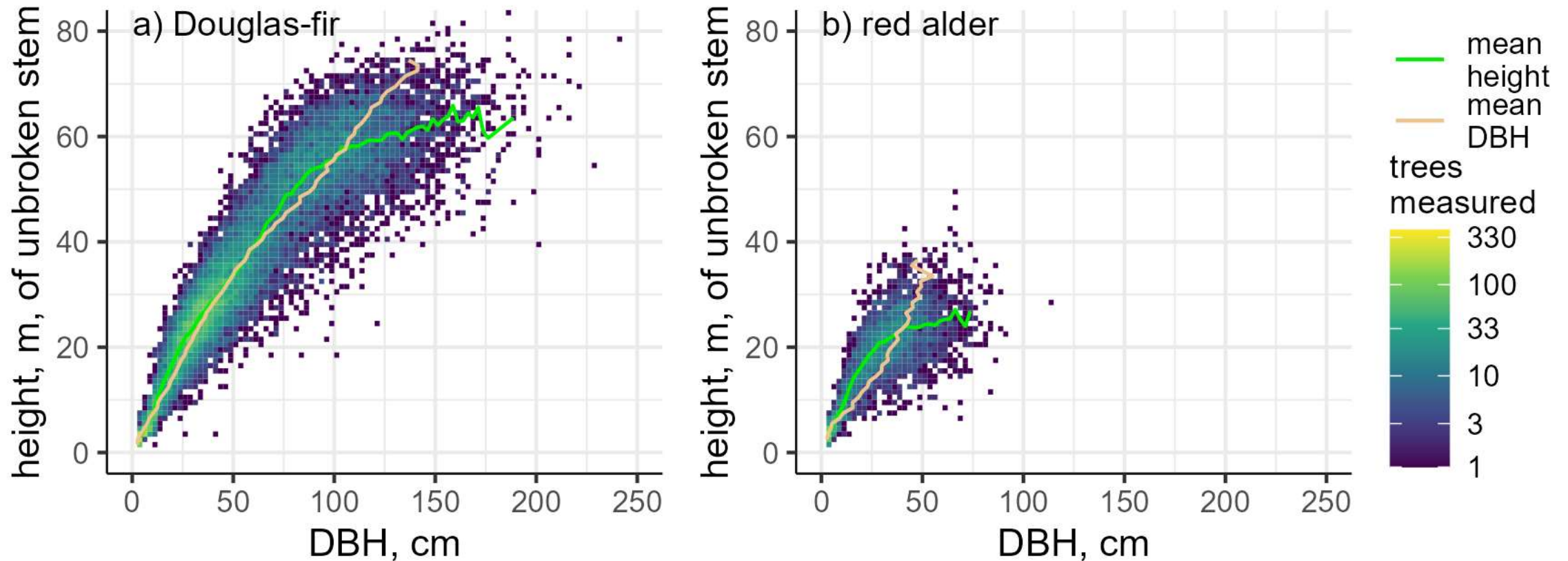
- Coastal rainforest
 - ~33,000 ha
 - 70–110% slopes
 - 1900 – 2800 mm H₂O yr⁻¹
 - T_{annual} 10–12 °C
- Complex history
 - Multifire stands (1800s)
 - Plantations (1950+)





Data

- 2016 ground inventory
 - 18,363 plots: 97,424 trees with 33,447 height measure trees





Models

- Nonlinear regression
 - 10 DBH forms generalized
 - 13 height forms generalized
- Generalized Additive Models (GAMs)
 - One form each for DBH and Height
- Linear regression
 - Linear and parabolic controls

$$DBH = a_1(H - 1.37)^{b_1} e^{b_2(H-1.37)}$$

$$H = 1.37 + a_1 H_{100}^{b_1} \left(1 - e^{b_2 \left(\frac{TPH}{BA} \right)^{b_3} DBH} \right)^{b_4}$$

$$DBH = a_0 + \sum_i a_i f_i(H)$$

$$DBH = a_1(H - 1.37)$$

$$DBH = a_1(H - 1.37) + a_2(H - 1.37)^2$$



GAMs [in this study]

- Penalized thin plate splines
 - Constrained flexibility
 - Fit using REML
- Nonparametric controls
 - Assess structural error
- Can be computationally intensive



Generalizing Predictors

- Stand history
 - natural regeneration, plantation
- Physiography
 - physio = elevation + slope + aspect + TSI
- Relative height and relative DBH
 - $RelHt = H / H_{100}$, $RelDbh = DBH / QMD$
- Basal area
 - $BA+L$ = basal area (BA) + basal area larger than current tree (BAL)
 - $ABA+T$ = approximate BA (ABA) + basal area taller (AAT)



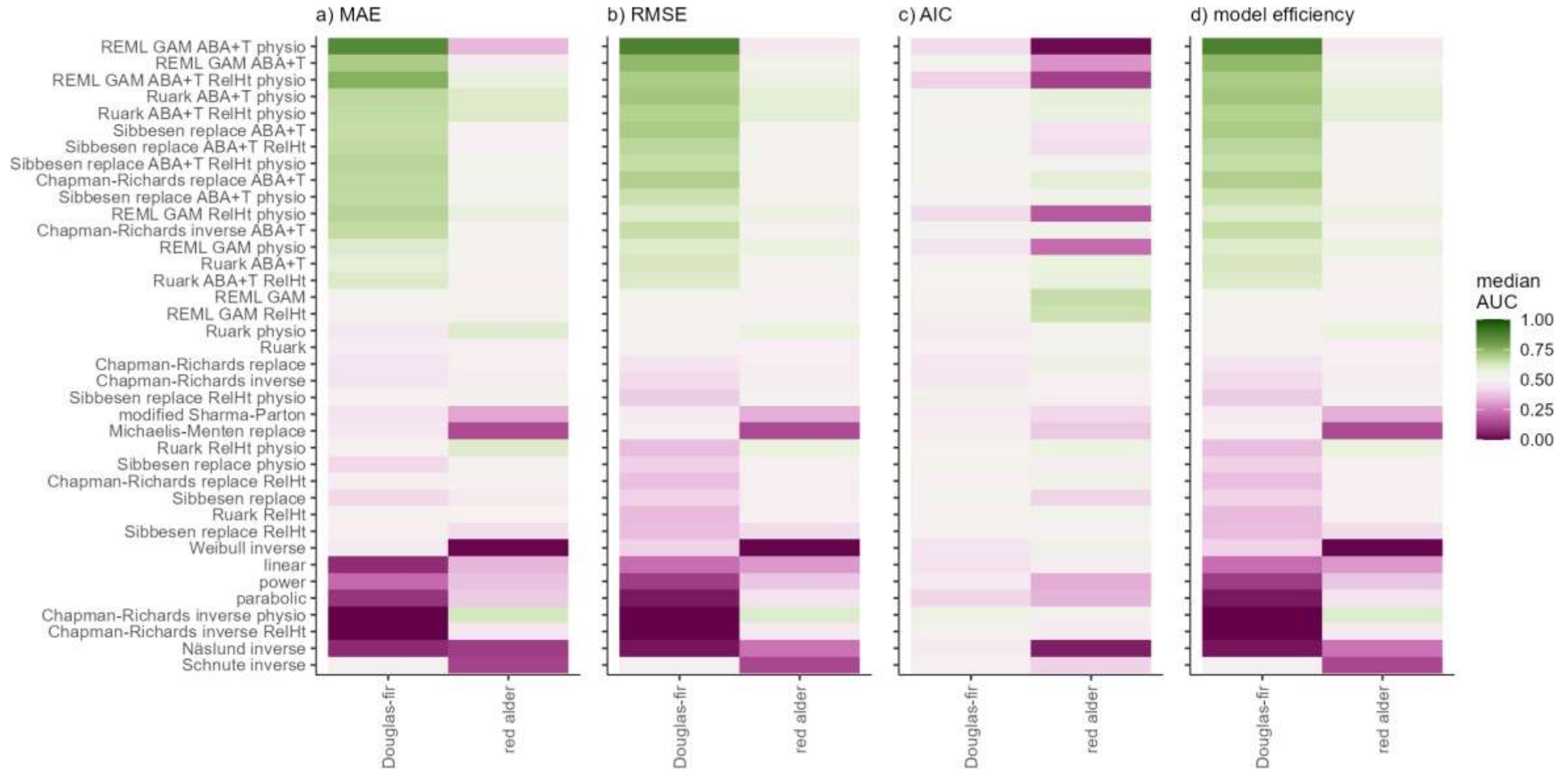
Model Assessment

- Cross validated ROC AUC
 - Rank models by predictive ability (0 = least accurate, 1 = most)
- Goodness of fit
 - MAE, RMSE, AIC, model efficiency
 - Nash-Sutcliffe definition

$$efficiency = 1 - \frac{\sum (Obs - Model)^2}{\sum (Obs - \overline{Obs})^2}$$

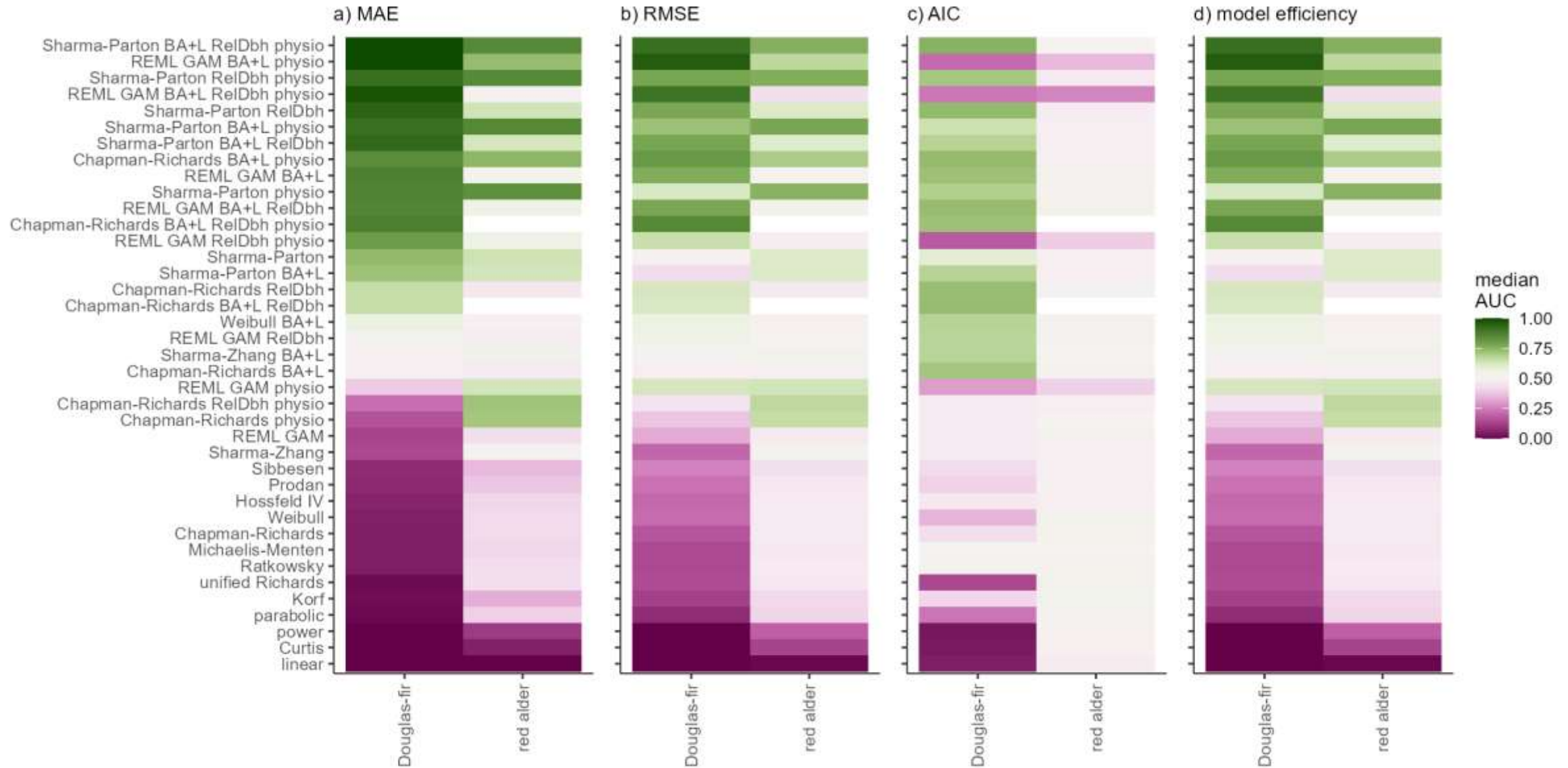


Results: DBH



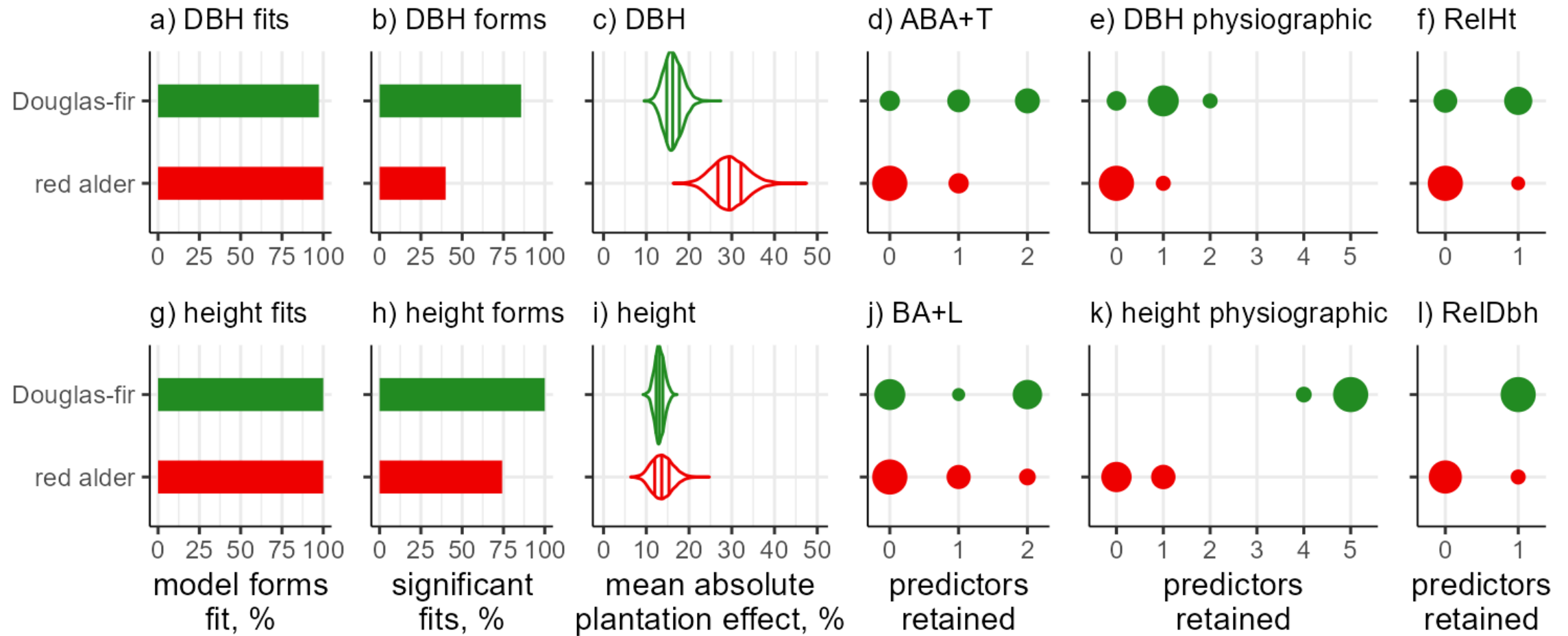


Results: Height





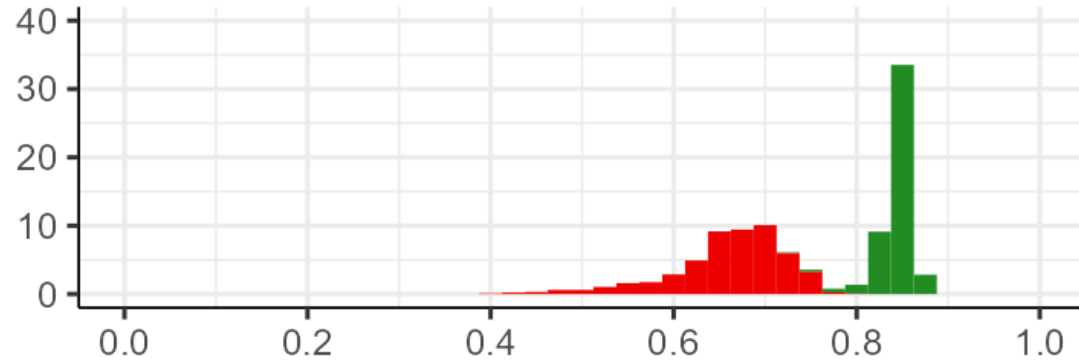
Results: Forms and Predictors



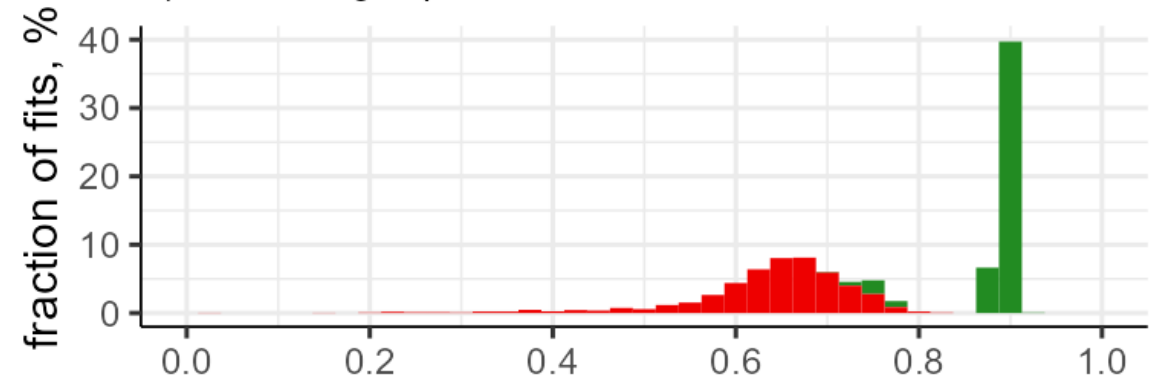


Results: Accuracy

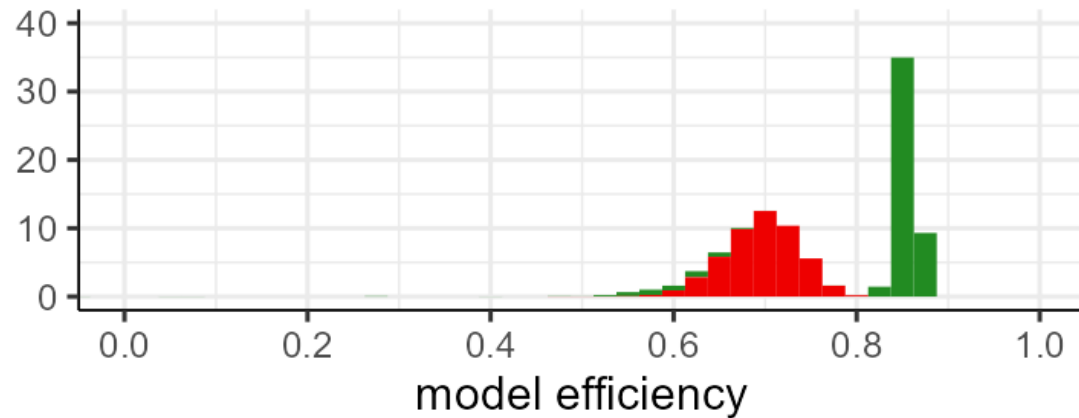
a) base DBH prediction



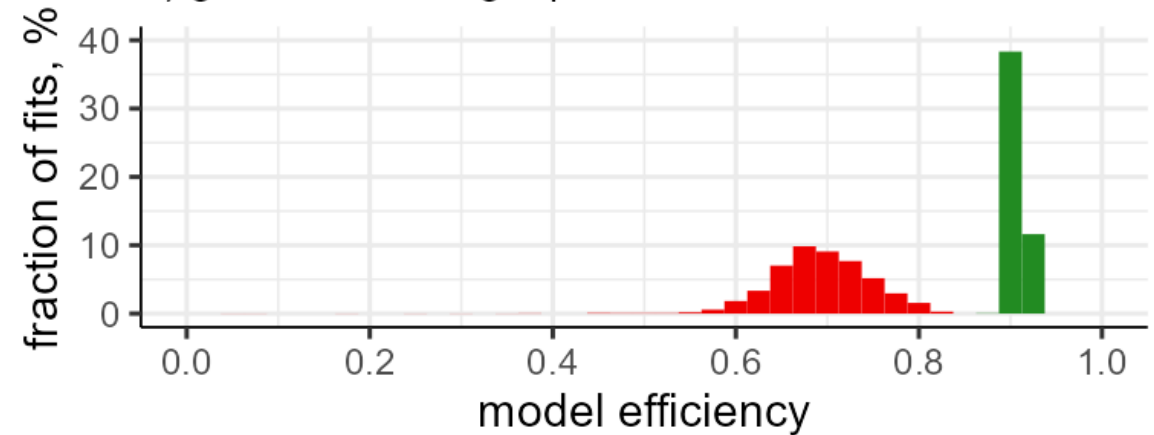
b) base height prediction



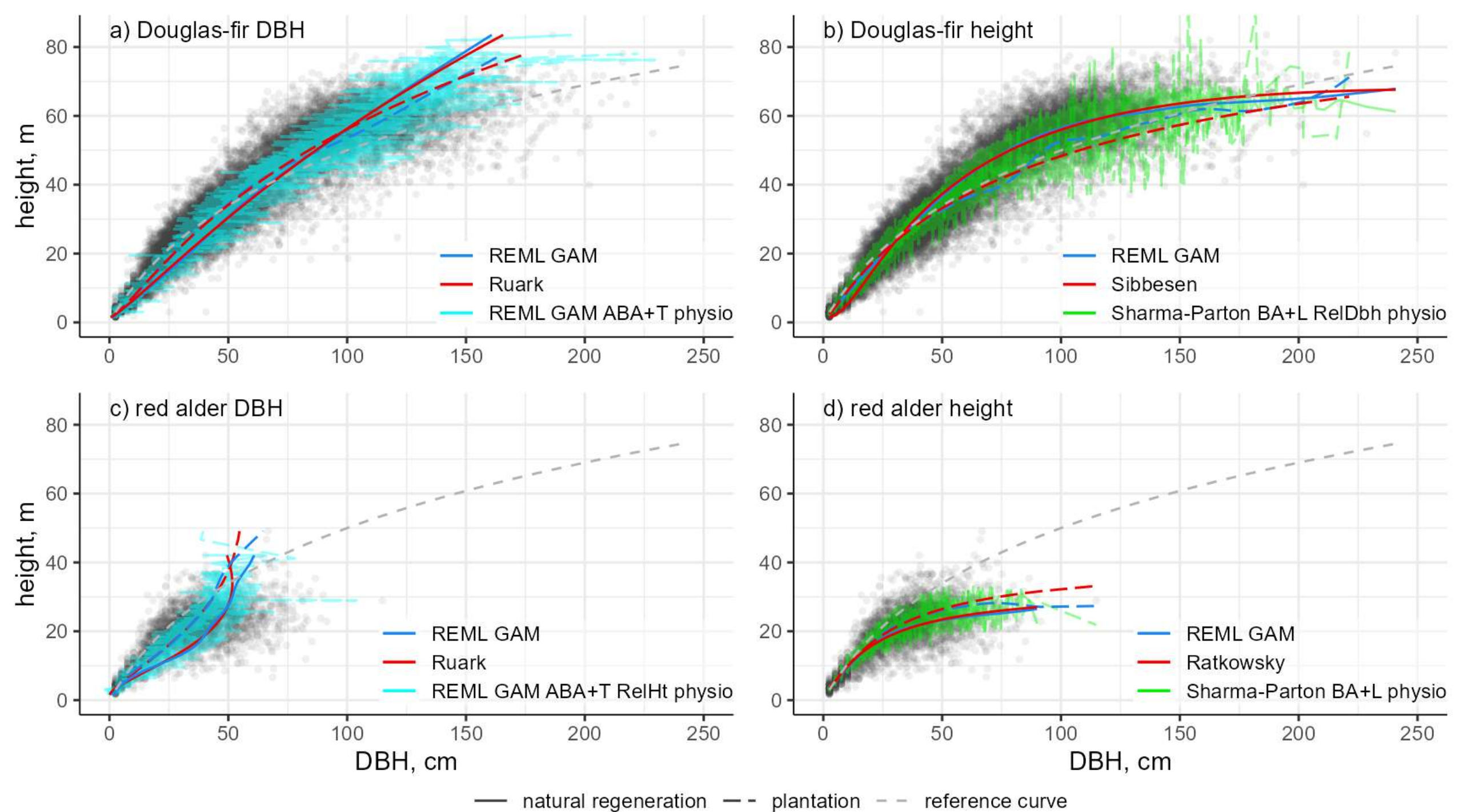
c) generalized DBH prediction



d) generalized height prediction



 Douglas-fir  red alder





Main Findings

- Species dependent model efficiency
 - > 0.85 for Douglas-fir
 - > 0.70 for red alder
- Nonlinear regression form
 - Height: Sharma-Parton
 - DBH: Ruark
- Prediction of 1 million trees
 - Nonlinear: < 1 sec
 - GAM: > 1 minute (base GAM)
 - > 10 minutes (complex GAM)



Conclusion

- Control forms are valuable
 - GAMs can be more accurate (not by much)
 - Simple linear regression sometimes wins (for other species)
- Chapman-Richards
 - Not as accurate as others
 - Harder to fit
 - The ONLY One Invertible: Height and DBH

$$H = 1.37 + a_1(1 - e^{b_1 DBH})^{b_2}$$

$$DBH = a_1 \log\left(1 - \min(b_1(H - 1.37)^{b_2}, l_1)\right)$$

$$DBH = a_1(1 - e^{b_1(H-1.37)})^{b_2}$$



Unexpected result

1. Invertibility is not consistent:
 - Functional or Mathematical

2. Complements the Incompleteness Theorem of Gödel
 - Not rooted in Aristotelian logic weakness: "I am lying"



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