

Can forest carbon be optimized following an eastern spruce budworm outbreak?

INTRODUCTION

Eastern spruce budworm

(Choristoneura fumiferana)

4.7 million ha of spruce-fir at risk of defoliation in northern New England and New York (2x state of NH)¹

Figure 1: Basal area at risk of defoliation from spruce budworm.³



Previous study by Gunn et al. 2020

- Widespread defoliation and management response (salvage dead wood) has implications for forest carbon
- In most cases, salvaging leads to net carbon emissions in years 0-20 after treatment
- Salvaging leads to a net carbon sequestration over longer time periods (years 20-40 after treatment)

RESEARCH QUESTIONS

1. Can we optimize forest carbon following a major disturbance using machine learning methods?

2. Are there stand conditions (e.g. basal area of balsam fir) where not salvaging may be preferred from a carbon perspective?

3. How does economic discounting affect carbon optimization results? Discounting captures a preference for near-term storage and penalizes longer wait periods before sequestering more carbon with salvage logging.

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Decision tree results

Latitude

- Decision trees identify key predictors to optimize forest carbon
- Tree structure varies based on (1) treatment scenario and (2) discount rate

Figure 4: (Right) Decision tree models produce trees to optimize carbon values at each plot.⁴ Branches are created from the most important predictors variables. Stand characteristics can be used to follow each branch to a decision of whether salvaging or not salvaging likely stores the most carbon at year 40.



Salvage

No Salvag





RESULTS



Figure 5: Decision tree branches (important predictor variables) change based on harvest scenario and discount rate scenario. In this figure a discount rate of 5% generates models with different tree branches compared to figure 4 that shows a discount rate of 1% scenario.

CONCLUSIONS

Stand characteristics:

Higher basal area of host species (balsam fir and spruce species) and larger host trees (higher QMD) could suggest not salvaging may be preferred

Higher discount rates (valuing near-term storage) suggests salvaging less often may be preferred

More volume removed from salvaging = greater swings in forest carbon budget

If there is a lot of wood salvaged, and you care about GHG emissions, choose to **not salvage or salvage less**

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