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# Formulating climate sensitive yield projections using transfer functions

D. Sattler<sup>a</sup>, K. Peterson<sup>b</sup>, T. Wang<sup>b</sup>, and G. O'Neill<sup>c</sup>

<sup>a</sup> Canadian Forest Service, Victoria
<sup>b</sup> University of British Columbia, Vancouver
<sup>c</sup> Government of British Columbia, Kalamalka Forestry Centre

## **Yield Projections**

- Silvicultural planning
- Supports Allowable Annual Cut (AAC) determinations
- Carbon Budget Monitoring
- National wood supply
- ...many other applications

## Yield Projection Systems and Climate

#### • Why the need for climate variables within G&Y models?

- Climate change is accelerating
- Longer projection periods
- Climate affects productivity (30% of variability in Site Index due to climate<sup>a</sup>)
- CBST
- Blue Ribbon Panel Report (2018); Penner Report (2021); Fletcher Report (2023)

### • BC models: climate static

- AB\*, SK, and MB models climate static (\*climate vars added to MGM)
- ON climate sensitive site index curves
- NB (NS, PEI) and QC some climate sensitive functions

<sup>a</sup> Monserud et al. 2006

## Yield Projection Systems in BC

#### **Unmanaged stands**







Stand-level Crown closure, Basal area, Site Index

Pure and mixed-species TSR and inventory updates Deterministic

#### Individual tree (TASS) Mixed-species (TASS III)

Meta-model (TIPSY) Stand-level (TIPSY) TSR (TIPSY)

#### Managed stands







TASS

TIPSY

Lovell

et

al.

N

023

## How to incorporate climate effects into G&Y?

#### 1. Temporary Sample Plots (Space-for-time substitution)

- Sample population = locally adapted trees
- Most studies predict increased site productivity
- Nigh et al. (2004) for site index [SI = f(Climate Vars.)]





## How to incorporate climate effects into G&Y?

#### 2. Permanent Sample Plots (time-for-time)

- Sample population = locally adapted stands
- "Marginal effect" of climate is usually small
- Oboite and Comeau (2020) for diameter growth, Cortini et al. (2017) for survival in MGM





## How to incorporate climate effects into G&Y?

## 3. Stem analysis

- Sample population = locally adapted trees
- Uncertainty around sample tree history
- Sharma et al. (2015) for Height-age (Site index) curves



## How to incorporate climate effects into G&Y?

- Approach 2 requires large extrapolation
- Future migration lag is underestimated
- Not designed to accommodate seed transfer
- Ignore genetic-byenvironment interaction



## Transfer Functions and Climate Change



(moved to colder)

## Transfer Functions and Climate Change

- TFs relate attributes (e.g., HT, Survival) to climate transfer distance
- Population-specific response to climate change
- Accounts for climate-maladaptation
- Less extrapolation
- How to link to G&Y models?
- Of limited use within TSR



## Transfer functions in Growth and Yield Modeling

O'Neill and Nigh (2011): Adjust Height-Age curves for Lodgepole Pine using Transfer Function for Relative Height

- Illingworth Data (est. 1974, 60 sites, 142 pops.)
- Tested in TASS

#### **Objectives:**

- 1. Revisit TF for Height
- 2. Formulate a TF for Survival
- 3. Link both the TF for Height and TF for survival to a GY model

![](_page_10_Figure_9.jpeg)

# Using TFs to modify gr

- Revised TF for Relative Height (Hp)
  - Includes a population-specific interaction term
  - **GLM** ( $Y \sim N(u, \sigma^2)$ ), where:  $(q(\mu) = ln(\mu))$
- Relative Height (Hp) used as a multiplier in Height-Age equations

![](_page_11_Figure_6.jpeg)

Site

# Using TFs to modify growth models

## • TF for Survival

- Dynamic over time
- Model fit using repeated measures from Illingworth Trial
- Genetic-by-environment interaction

#### • Tree Survival in TASS

- Mortality controlled by neighbourhood, crown size, random events
- Too complex

![](_page_12_Figure_9.jpeg)

![](_page_13_Figure_1.jpeg)

![](_page_14_Figure_0.jpeg)

## Simulations at three cutblocks (local seed)

![](_page_15_Picture_1.jpeg)

17-22% reduction in stand volume

![](_page_15_Figure_3.jpeg)

## Limitations

- Provenance Trials ≠ G&Y Plots
  - No buffers
  - Single planting density
  - < 18 trees / population</p>
  - No crown measurements
  - Trials << Rotation Age
- Transfer functions for all species?
  - TFs for Generalists and Specialists?

# Next Steps

- Simulations with non-local seed
- Simulations with local/nonlocal seed mixtures
- Landscape-level analysis harvest scheduler
- Adding the option to modify projections in TIPSY???

# Simulating the impacts of pine rusts

![](_page_17_Picture_1.jpeg)

![](_page_18_Figure_0.jpeg)

## Data for model building

## Main Data Sources:

- EP 671 (1 installation; Est. 1968)
  - 4 planting densities, each replicated in 6 plots (24 plots)
  - 64 trees/plot (+ buffers) measured at 8, 13, 18, 23, 28, 33, and 41 years
  - Rusts surveys performed in 1997, 2002, 2005, 2014

#### • EP 964 (5 installations; Est. 1987-1991)

- 5 planting densities, each replicated in 3 plots
- 64 trees/plot measured every 5 or 10 years, up to 25 years
- Rust survey performed in 2016
- Bednesti (1 installation; Est. 1988)
  - Various planting densities + treatments over 30 plots
  - Most plots measured annually up to 30 years
  - Rusts surveyed in 2010, 2011, 2015, 2016, 2017

![](_page_19_Picture_14.jpeg)

![](_page_19_Figure_15.jpeg)

## Heights of gall on main stem

![](_page_20_Figure_2.jpeg)

## **Cumulative incidence of first infection**

![](_page_21_Figure_2.jpeg)

![](_page_21_Figure_3.jpeg)

![](_page_21_Figure_4.jpeg)

## **Cumulative incidence of first infection**

## **Weibull Growth Function** 0.6 Cumulative\_Inc = $\beta_1 [1 - \exp(\beta_2 t^{\beta_3})]$ 0.05 0.05 Initialized using Free Growing Survey Data • $\beta_1$ = initialization variable 0.04 Annual probability of infection **Differential form (annual probability):** $P_{inf} = \beta_1 \beta_2 \beta_3 t^{\beta_3 - 1} \left[ \exp(-\beta_2 t^{\beta_3}) \right]$ 0.01

![](_page_22_Figure_3.jpeg)

## Modified probability of infection

Two parameter Weibull:

$$CAP_1 = 1 - \exp\left(\left[-\left(\frac{RelH}{\alpha_1}\right)^{\alpha_2}\right]$$

#### **Inverse transform:**

$$RelH_u = \alpha_1 [-\ln(1-u)]^{1/\alpha_2}$$

(maps U(0, 1) to Weibull distribution)

- Draw u~U(0, 1), generate  $RelH_u$
- Infect trees where:  $RelH_i < RelH_u$ until we reach  $P_{inf}$  for that year

![](_page_23_Figure_9.jpeg)

2500/ha planted, 20% incidence, standard juvenile mortality

![](_page_24_Picture_2.jpeg)

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## Galls >50% encirclement

![](_page_25_Figure_2.jpeg)

![](_page_25_Picture_3.jpeg)

**Rust module** Infect trees Adjust growth of infected trees? Adjust probability of mortality Adjust log bucking

- Is a Bernoulli random variable (either 0 or 1)
- If > 50% = 1, otherwise 0

## Expected Loss in (A) Merch. Vol. and (B) Lumber

![](_page_26_Figure_2.jpeg)

## Questions?

![](_page_27_Picture_2.jpeg)

![](_page_28_Figure_1.jpeg)