



Silviculture Note #??

Stocking estimators and future volume

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September 4, 2002

1. Introduction

Tree stocking, the degree to which growing space is occupied, is an important forestry concept. In a young stand, stocking is a main determinant of future stand volume/ha and the distribution of tree sizes at harvest (Clutter et al. 1983). A variety of estimators have been developed to quantify the stocking in regenerated stands (Stein 1978; Shreuder, Gregoire and Woods 1993, pg 291). Typically, these estimators are designed to take on greater values as density and uniformity of tree distribution increase, and many are capped at some maximum value. When stocking estimators are evaluated, the focus has been on ease of use, cost, and the extent to which they exhibit desired behaviour over a range of tree density and uniformity (Stein 1978). Historically, the ability of a stocking estimator to predict future volume has not been an important evaluation criterion.

Recently in B.C. interest has grown in the relationship between stocking estimators and future volume. Bergerud (2001) demonstrated the relationship between the stocking estimator "total well-spaced trees/ha" and TASS predicted merchantable volume/ha at age 67 years for lodgepole pine on site index 18 m. J.S. Thrower and Associates (2002) developed a new stocking estimator "MSQ" and demonstrated its relationship to TASS predicted merchantable volume/ha at age 80 years for lodgepole pine on site index 20 m.

Martin, Browne-Clayton, and McWilliams (2002) described a new system for managing reforestation that is based on the future volume predicted, in part, from the stocking observed in young stands. Though this new system uses the stocking estimator "MSQ," it could be re-formulated to use other stocking estimators. In future implementations of this new system, it is desirable to ensure that the stocking estimator used has high predictive power. In this paper, I report the results of a cursory assessment of the ability of four stocking estimators to predict merchantable volume/ha at age 80 years for lodgepole pine on site index 18 m.

2. Methods

The spatially explicit, individual tree growth model TASS (Mitchell 1975, Mitchell and Cameron 1985) was used to generate a variety of tree spatial patterns in a 100 m x 100 m plot. From bare ground the stand represented by each plot was grown to the silviculture survey date at which time surveys were simulated in the stand. The survey parameters were computed and the stand was grown for 100 years. Volumes



at ages 60, 80, and 100 years (site heights of 18.8, 21.9, and 24.0 m, respectively) were extracted from the TASS output, though only the volume at age 80 is reported here. Regression analysis was used to assess the strength of the relationship between the four stocking estimators and merchantable volume/ha at age 80.

2.1 Stocking estimators

Though a total of seven stocking estimators were evaluated, in this paper I report only the results for four (Table 1):

Table 1. Description of the four stocking estimators that were assessed.

| Code | Name | Plot procedure | Compilation |
|-------|--|---|--|
| ТТРН | Total trees per hectare | In a 3.99 m radius plot, the surveyor counts all live trees. | Plot counts are averaged and expanded to a per hectare basis. |
| WSTPH | Well-spaced trees per hectare | In a 3.99 m radius plot, the surveyor maximizes the count of well-spaced trees. No "M" cap. 2.0 m MITD. | Plot counts are averaged and expanded to a per hectare basis. |
| MSQ | Mean stocked quadrants | In a 3.99 m plot divided into quarters along cardinal directions, the surveyor counts the number of quarters containing at least one live tree. | Plot counts are averaged. |
| PERSP | Percent stocked 1.4 m radius plots | The surveyor counts a 1.4 m plot as stocked if it contains at least one live tree. | Percent of all plots that were tallied as stocked is computed. |

2.2 TASS simulations

Fifty different tree spatial distributions were taken from the many stem maps used to produce Land Management Handbook 50 (Bergerud 2002). From those distributions classified as clumped, maps with the following initial trees/ha were used: 300, 425, 550, 650, 750, 900, 950, 1020, 1150, 1240, 1400, 1500, 1750, 2000, 2250, 2500, 2750, 2900, 3100, 3265, 3906, 4500, 5200, 5917, 6944, 8000, 10000, and 20000. From those distributions classified as natural (random spatial pattern), maps with the following initial trees/ha were used: 300, 550, 750, 950, 1150, 1400, 1750, 2250, 2750, 3100, 3906, 5200, 6944, and 10000. From those distributions classified as planted (grid spatial pattern), maps with the following initial trees/ha were used: 425, 650, 950, 1240, 1750, 2500, 4500, and 8000.

The following run specifications were used for each TASS simulation:

TASS version: v2.07.14WS

Species: interior lodgepole pine

Site index: 18 m

Site index curve code: PI THROWNIGH

Merchantable volume

Minimum dbh: 12.5 cm



Top dib: 10 cm

Stump height: 0.3 m

OAFs: No OAFs applied

Plot size: 100 m X 100 m

The TASS runs and the survey simulations were conducted by RamSOFT Systems Ltd.

2.3 Survey simulation

Each stem map was grown to a site height of 5 m, which occurred 16 years from run initialization. Surveys were simulated at this time. Ten plots were randomly located on the stem map, plot values taken, and the sample mean computed. This was repeated 1000 times. Last, the 1000 sample means were averaged. Thus, each survey value is a mean from 10,000 plots. In counting trees, no minimum height criteria were applied. To reduce costs by re-using data previously compiled, one set of plot centers was used for WSTPH and MSQ and a different set for the other estimators.

2.4 Data analysis

A single equation form was identified that could provide a good fit to each of the four volume-stocking estimator relationships. A function in the Weibull family was fit with nonlinear least squares using the SYSTAT statistical software (SPSS Inc. 1998):

$$V = b_0 \left(1 - \exp \left(b_1 \left(\frac{X_{\phi}}{\phi} \right)^{b_2} \right) \right)$$

Where V is merchantable volume/ha at age 80,

 b_0 , b_1 , and b_2 are parameters, and

X is the stocking estimator (TPH, WSTPH, MSQ, and PERSP). ϕ is a constant assigned before fitting equal to the largest X value in the data set: $\phi = 16712$ for TPH, $\phi = 2182$ for WSTPH, $\phi = 4$ for MSQ, and $\phi = 100$ for PERSP.

The fit statistics and a visual examination of residuals indicated that excellent fits were obtained. Two fit statistics, the mean square error and the squared correlation between observed and predicted values, were taken to indicate the ability of a stocking estimator to predict future volume/ha (Table 2).

The complete data set is provided in Appendix A.

3. Results

The relationship between each stocking estimator and TASS-predicted volume at age 80, with the fitted curve, is displayed in Figures 1-4.



Though volume/ha at 80 years is approximately linearly related to MSQ, the relationship is curvilinear with TTPH, WSTPH, and PERSP. A visual assessment suggests that the stocking estimators TPH, WSTPH and PERSP produce values that are spread more widely, while many of the 50 stem maps assessed returned MSQ values very close to 4. However, an increased spread is not associated with an improved ability to predict future volume (Table 2).

MSQ predicts future volume/ha slightly better than WSTPH and PERSP do and much better than TTPH does (Table 2). The relationships between future volume and WSTPH, PERSQ, and MSQ are so strong that little improvement can be expected from adding additional explanatory variables or stratifying the data.

Table 2. Fit statistics from regressions relating stocking estimators to future volume.

| Stocking estimator | Mean square error | R ² : Correlation of observed and predicted values (squared) |
|--------------------|-------------------|---|
| ТТРН | 703 | 0.84 |
| WSTPH | 142 | 0.97 |
| MSQ | 44 | 0.99 |
| PERSP | 152 | 0.97 |



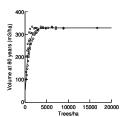


Figure 1. Relationship between merchantable volume/ha at age 80 and total trees/ha at survey. Solid line is fitted regression.

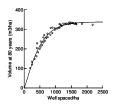


Figure 2. Relationship between merchantable volume/ha at age 80 and total well spaced trees/ha at survey. Solid line is fitted regression.

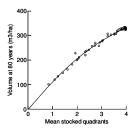


Figure 3. Relationship between merchantable volume/ha at age 80 and mean stocked quadrants at survey. Solid line is fitted regression.

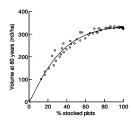


Figure 4. Relationship between merchantable volume/ha at age 80 and percent stocked plots at survey. Solid line is fitted regression.



4. Discussion

In terms of its ability to predict future volume, the stocking estimator MSQ is superior to the three others examined (TTPH, WSTPH, and PERSP). However, WSTPH and PERSP are almost as good so when other factors are considered, such as cost, ease of use, or familiarity of procedure – the use of MSQ, WSTPH, or PERSP could easily be justified. As expected, TTPH is a poor estimator of future volume and, for this purpose, its use is not recommended.

Bergerud (2001) found that the relationship between TASS-predicted future volume and well-spaced trees/ha varied with tree spatial pattern. Though this issue was not specifically examined in this analysis, the excellent fits suggest that little could be gained by stratifying by tree distribution type (random, clumped, or grid). Additional replicates of the planted and natural spatial pattern stem maps and subsequent analysis is recommended to further examine this issue.

The volumes used in this study are TASS-predicted volumes, not actual volumes observed in real stands that originated with the specified tree spatial patterns. Thus, the fit statistics grossly over-state the accuracy with which these stocking estimators will predict real stand future volumes. Moreover, if there is some systematic bias in TASS predictions, for example, if volumes are consistently over-estimated at low stockings, then the shape of the volume-stocking relationships displayed in Figure 1-4 will be incorrect.

These results indicate the correlation between future volume and a stocking estimator when sample size is enormous. Each data point is the mean of 10,000 sample plots. The correlation under operationally realistic sample sizes should be investigated. Furthermore, it would certainly cost less to take a single PERSP plot than to take a single WSTPH plot. In dense stands, TPH is also time consuming to tally. However, cost has not been considered in this analysis. Subsequent study should attempt to identify the stocking estimator that provides the most accurate prediction of future volume at a realistic fixed cost.

5. Literature cited

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Appendix A: Complete data set

| Stem map # | TASS initialization density (#/ha) | Tree spatial distribution type | Trees per hectare at survey (#/ha) | Well-spaced trees per hectare (#/ha) | Mean stocked quadrants with unrotated 3.99 m plots | Percent stocked 1.4 m radius plots (percent) | Merch. volume at age 80 (m3/ha) |
|---------------|---|--------------------------------|---|---|---|---|---|
| 1 | 300 | clumped | 278 | 192 | 0.84 | 12.81 | 101 |
| 2 | 425 | clumped | 402 | 254 | 1.09 | 16.64 | 120 |
| 3 | 550 | clumped | 524 | 334 | 1.42 | 21.28 | 148 |
| 3 4 | 650 | clumped | 628 | 394 | 1.61 | 26.23 | 162 |
| 5 | 750 | clumped | 716 | 446 | 1.80 | 28.50 | 182 |
| 6 7 | 900 | clumped | 852 | 500 | 1.99 | 32.54 | 200 |
| | 950 | clumped | 898 | 528 | 2.08 | 33.99 | 211 |
| 8 | 1020 | clumped | 948 | 576 | 2.34 | 36.91 | 220 |
| 9 | 1150 | clumped | 1086 | 616 | 2.33 | 39.68 | 233 |
| 10 | 1240 | clumped | 1170 | 662 | 2.48 | 43.32 | 241 |
| 11 | 1400 | clumped | 1290 | 706 | 2.61 | 44.67 | 256 |
| 12 | 1500 | clumped | 1400 | 752 | 2.72 | 48.21 | 259 |
| 13 | 1750 | clumped | 1658 | 836 | 2.93 | 54.34 | 272 |
| 14 | 2000 | clumped | 1908 | 896 | 3.05 | 59.73 | 281 |
| 15 | 2250 | clumped | 2156 | 978 | 3.25 | 63.15 | 297 |
| 16 | 2500 | clumped | 2314 | 1046 | 3.38 | 67.82 | 303 |
| 17 | 2750 | clumped | 2582 | 1104 | 3.52 | 71.72 | 307 |
| 18 | 2900 | clumped | 2670 | 1128 | 3.58 | 72.64 | 308 |
| 19 | 3100 | clumped | 2894 | 1168 | 3.65 | 75.81 | 320 |
| 20 | 3265 | clumped | 3064 | 1196 | 3.63 | 76.23 | 317 |
| 21 | 3906 | clumped | 3676 | 1306 | 3.79 | 83.15 | 328 |
| 22 | 4500 | clumped | 4234 | 1356 | 3.84 | 86.54 | 329 |
| 23 | 5200 | clumped | 4842 | 1416 | 3.92 | 90.25 | 326 |
| 24 | 5917 | clumped | 5350 | 1468 | 3.95 | 93.00 | 329 |
| 25 | 6944 | clumped | 6218 | 1512 | 3.96 | 95.21 | 333 |
| 26 | 8000 | clumped | 7184 | 1572 | 3.98 | 97.32 | 331 |
| 27 | 10000 | clumped | 8892 | 1648 1840 | 3.98 | 98.58 | 328 |
| 28 | 20000 | clumped | 16712 | 250 | 3.98 1.22 | 99.95 | 328 |
| 29 30 | 300 550 | random random | 286 518 | 250 414 | 1.93 | 16.76 27.34 | 134 228 |
| 31 | 750 | random | 702 | 530 | 2.38 | 34.83 | 237 |
| 32 | 950 | random | 890 | 640 | 2.36 2.74 | 43.77 | 270 |
| 33 | 1150 | random | 1086 | 742 | 3.02 | 50.62 | 277 |
| 34 | 1400 | random | 1310 | 846 | 3.25 | 56.31 | 307 |
| 35 | 1750 | random | 1642 | 976 | 3.52 | 64.82 | 305 |
| 36 | 2250 | random | 2102 | 1124 | 3.73 | 73.93 | 326 |
| 37 | 2750 | random | 2578 | 1248 | 3.85 | 81.13 | 328 |
| 38 | 3100 | random | 2910 | 1322 | 3.90 | 85.30 | 326 |
| 39 | 3906 | random | 3662 | 1432 | 3.96 | 90.55 | 334 |
| 40 | 5200 | random | 4810 | 1474 | 3.97 | 95.25 | 329 |
| 41 | 6944 | random | 6332 | 1546 | 3.98 | 98.17 | 331 |
| 42 | 10000 | random | 8906 | 1660 | 3.98 | 99.79 | 325 |
| 43 | 425 | planted | 406 | 406 | 2.03 | 24.58 | 204 |
| 44 | 650 | planted | 612 | 612 | 2.89 | 36.40 | 258 |
| 45 | 950 | planted | 906 | 904 | 3.59 | 54.18 | 308 |
| 46 | 1240 | planted | 1170 | 1158 | 3.82 | 69.25 | 324 |
| 47 | 1750 | planted | 1646 | 1472 | 3.92 | 86.22 | 334 |
| 48 | 2500 | planted | 2336 | 1594 | 3.98 | 95.90 | 328 |
| 49 | 4500 | planted | 4150 | 1644 | 3.98 | 99.55 | 331 |
| 50 | 8000 | planted | 7138 | 2182 | 3.98 | 100.00 | 323 |

Sustainable Forest Management Plan

