

1 **The Effect of Hardwood Component on Grapple Skidder and**
2 **Stroke Delimber Idle Time and Productivity – An Agent Based Model**

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4 **1. ABSTRACT**

5 The forest industry is a capital intensive business and therefore high efficiency in management and
6 forest operations is a must. Maine has millions of acres of forest stands with tree diameters smaller than
7 30 cm. The harvest productivity in such stands is low compared to stands with larger diameter trees. A
8 recent harvest productivity study in Maine identified operational constraints for whole tree harvest
9 systems, but efforts to improve active operations would be expensive and time consuming. A common
10 practice to reduce costs and time consumption is to develop simulation models and implement new
11 ideas within them. We developed a production efficiency model that leverages an agent-based modeling
12 approach. The model is based on the interaction of two common forest machines (grapple skidder and
13 stroke delimber) and incorporates empirical cycle time estimates from research in Maine. Four scenarios
14 have been developed to investigate baseline conditions, two GPS/GIS improvements, and the use of two
15 grapple skidders. The goal of this paper is to document a new agent based model and to investigate the
16 effect of hardwood component on machine idle time and productivity. Results showed that system
17 productivity was affected by skidding distance, bunch spacing, and removal intensity. An increase in
18 hardwood component led to a decrease in stroke delimber idle time but did not affect grapple skidder
19 idle. Further, hardwood component did not affect system productivity, and none of the three single-
20 skidder scenarios tested performed any better than another. We validated the model by conducting a
21 sensitivity analysis to confirm previous research results. The modeled waiting times are well within the
22 range of observed values and therefore suggest that this model is accurate and well calibrated. Our
23 conclusions are that when operating under average harvesting conditions there is no loss in productivity
24 due to a change in hardwood component and that a stroke delimber idle time of 40% or more is
25 unavoidable unless the stroke delimber can work independently. Future applications of this model may
26 target specific production forestry conditions. Suggested analyses include productivity gains from
27 technological improvements as well as the unit cost of production under a variety of stand and site
28 conditions.

29

30 2. INTRODUCTION

31 Due in part to regenerating clearcuts from the spruce budworm era in the 1970s and 1980s, forest
32 operations managers in Maine must manage an increasing percentage of stands that consist of small-
33 diameter stems (dbh <30 cm). Approximately 11 million acres of forest land in Maine contain or are
34 dominated by trees smaller than 30 cm in dbh (McCaskill et al.). Forest operations are an important part
35 of the forest industry but are also very capital intensive (Purfürst). Due to the high capital investment in
36 harvesting equipment, and the cost of running the machines, it is important to know machine
37 productivity to fully utilize the individual machines. Effective management of forest operations therefore
38 requires accurate estimates of harvest costs and productivity, although the monitoring of these
39 variables may be difficult (Wang et al.; Holzleitner et al.). The two dominant and fully mechanized
40 harvesting systems in Maine are whole-tree (feller-buncher, grapple skidder, stroke delimber) and cut-
41 to-length (harvester, forwarder) (Leon and Benjamin). As the names suggest, whole-tree harvesting
42 operations severe the tree from the stump and then transport it to the roadside including all branches
43 as a whole tree, cut-to-length harvesting operations on the other side, severe trees from the stump and
44 then cut off the branches and crosscut the bole into specific length logs, which are then transported to
45 the roadside. These harvesting systems are generally very productive when operating in large diameter
46 tree stands but have a reduced productivity when operating in small diameter tree stands (P Hiesl and
47 Benjamin). With high investments in equipment it is therefore crucial to achieve high machine
48 productivities to keep the unit cost at a low level. To increase the productivity of individual machines
49 and the harvesting system it is therefore necessary to improve or change existing harvesting processes.

50 The primary goal of any logging contractor is to generate revenue to pay for the equipment and to
51 create income. Maximizing machine utilization is one way to reach this goal and is described as the most
52 important objective of a logging contractor (Bolding). In order to maximize the utilization of a machine it
53 is important to know where bottlenecks are. Several methods are available to identify these
54 bottlenecks. Time studies are a common tool to evaluate harvesting operations and identify bottlenecks,
55 however, they can be rather time consuming (Bolding; Bazghandi; Bradley et al.). Another accepted
56 method to analyze the productivity and impact of a harvesting system are simulation models (Polley;
57 Baumgras et al.; Goulet et al.; Garner; Bradley et al.; Wang and LeDoux; Y. Li et al.). Also often used are
58 individual tree growth simulators such as Forest Vegetation Simulator (FVS) (Dixon), and regional
59 volume and taper equations (e.g. Li et al. 2012; Weiskittel and Li 2012). Individual tree growth models
60 are especially useful in combination with cycle time equations for harvesting equipment that are based
61 on individual trees (e.g. Hiesl and Benjamin 2015; Hiesl and Benjamin 2013a; Spinelli et al. 2010;
62 Adebayo et al. 2007). Simulation models have several benefits compared to time and motion studies,
63 such as fast execution of models and the possibility of changing system settings without changing the
64 real system (Polley; Bazghandi; Bradley et al.). The use of simulation models is not new to the forestry
65 community as simulation models have been available since the 1960's (Polley; Goulet et al.).

66 Before 2013, no harvesting productivity studies were conducted in Maine and therefore no up-to-date
67 productivity information for harvesting systems operating in Maine's forests was available to conduct
68 such computer simulations (P Hiesl and Benjamin). In 2012 and 2013, researchers at the University of
69 Maine collected cycle time and productivity data for five pieces of equipment (feller-buncher, harvester,
70 grapple skidder, forwarder, stroke delimber) commonly used in Maine, and developed cycle time and
71 productivity equations (Hiesl; P Hiesl and Benjamin; Patrick Hiesl and Benjamin; P Hiesl and Benjamin).
72 With these newly developed equations it is now possible to simulate the time consumption and
73 productivity of different harvesting systems in a variety of site and stand conditions. The logical
74 extension of the time and motion study conducted by Hiesl (2013) therefore is to use this data to
75 identify bottlenecks and to develop simulations with the new productivity data to test various scenarios

76 of possible improvements in forest operations. Observations during the field study showed that
77 harvesting operations consist of a large amount of non-productive waiting time.

78 Harvesting equipment used in whole-tree and cut-to-length harvesting systems mostly operate
79 independent from each other. The interactions between stroke delimiters (Figure 1) and grapple
80 skidders (Figure 2) are an exception to this. The grapple skidder delivers wood to be processed by the
81 stroke delimiter and often has to wait for the stroke delimiter to finish processing wood from the
82 previous load. Polley (1987) found that waiting times between 20 and 40% have to be expected due to
83 this dependency. The recommendation from Polley's research was to avoid such technological coupling
84 of new equipment. Today, however, these two machines are still very much dependent on each other.
85 Huth et al. (2004) commented that the existence of harvesting systems for many years and decades
86 does not necessarily mean that their use is sustainable. Today with decreasing profit margins, large
87 percentages of idle time due to technical coupling of grapple skidder and stroke delimiter cannot be
88 tolerated.

89 Whole-tree harvesting systems are the most important harvesting systems in Maine in terms of volume
90 cut (Leon and Benjamin). Unpublished data of Hiesl (2013) showed that there is a waiting time ranging
91 from 0% to 57% present when a grapple skidder and stroke delimiter work together at a variety of
92 commonly encountered site and stand conditions in Maine (Figure 3). With the feller-buncher working
93 independently we can therefore identify the interaction of the grapple skidder and the stroke delimiter
94 as the bottleneck in most whole-tree harvesting systems. Research has further shown that the
95 processing time of stroke delimiters is negatively impacted when processing hardwoods (Hiesl). As with
96 harvesters (Glöde), the generally larger branch size increases the processing time for stroke delimiters as
97 well. Maine's forest land consists of over 50% of hardwood forest types (McCaskill et al.), and land
98 managers and logging contractors alike have to deal with the negative impact of hardwoods on
99 harvesting productivity. This highlights the importance to understand the impact of an increasing
100 hardwood component on stroke delimiter and grapple skidder idle time.



101
102 Figure 1: A stroke delimiter generally processes one tree at a time by cutting of branches and the top
103 above a specific merchantable diameter.

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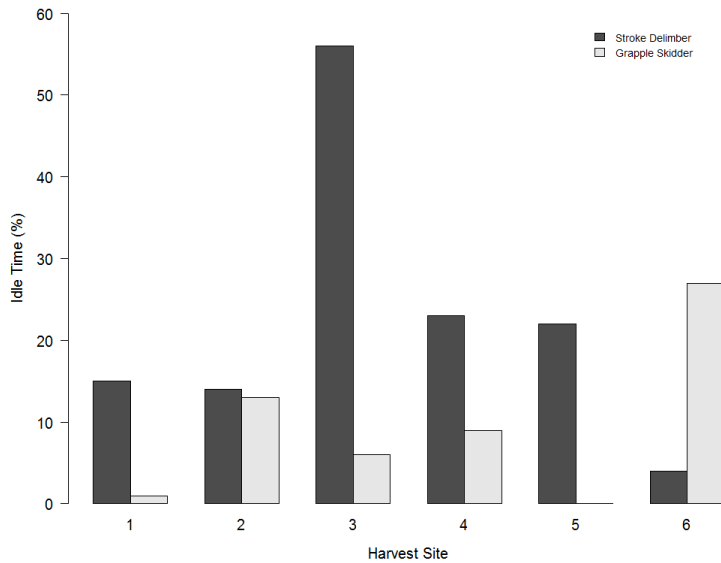


105
 106 Figure 2: A grapple skidder generally transports several trees in a bunch from the forest to the roadside
 107 where the whole trees get processed by a stroke delimeter.

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110 Figure 3: Observed percent idle time for stroke delimeter and grapple skidder from six different harvest
 111 sites observed during the summer of 2012 by Hiesl (2013). The sites represent a common range of site
 112 and stand conditions in Maine.



113

114 Presently, there are three computer simulation methods available for modeling different abstraction
 115 levels, such as System Dynamics, Discrete Event, and Agent Based (Borshchev and Filippov). All methods
 116 have their strengths and weaknesses, and we therefore focus on Agent Based Modeling (ABM) only.
 117 ABM is versatile and can be used in a range of low to high abstraction levels, depending on the needs of
 118 the simulation. With ABM the focus is on individual objects (agents) that can vary in their scope and
 119 nature, such as people, vehicles, machines, customers, competing companies, etc. (Borshchev and

120 Filippov). The novel aspect of ABM is that behavior rules of individual agents and their interactions can
 121 be specified. This is the most outstanding difference of ABM from the other simulation methods, and
 122 makes this method especially useful in modelling forest harvesting with different machines. We have
 123 chosen an agent-based modeling technique because we are focused on individual agents (stroke
 124 delimeter and grapple skidder) with unique and interacting behaviors. Epstein (1999) described five
 125 characteristics of agent based modeling (ABM) that aid in the decision making of the applicability of
 126 ABM to a certain research question: (1) heterogeneity of the agents, (2) autonomy of the agents, (3)
 127 explicit space, (4) local interactions, and (5) bounded rationality. All five characteristics hold true in our
 128 simulation of stroke delimeter and grapple skidder interactions. Although often used in social sciences,
 129 agent based modeling (ABM) experiences widespread popularity among other disciplines (Manson;
 130 Bazghandi; Gilbert). There is growing interest among researchers in using agent based models to explore
 131 ecological and silvicultural consequences of harvesting prescriptions (Arii et al.) and to investigate the
 132 harvest decision making of forest landowners (Leahy et al.). Our model will expand the use of agent
 133 based models to include forest operations research questions at the machine level.

134 Due to the large amount of data generated by this model and the multitude of research questions that
 135 can be asked we will focus in this paper on a detailed model description and investigate the effect of an
 136 increasing hardwood component on stroke delimeter and grapple skidder idle time and productivity. The
 137 analysis shown in this paper focuses on the hardwood aspect only, as a separate analysis of skidding
 138 distance, payload, and a two skidder scenario is part of a different article (Hiesl et al., submitted).

139

140 **3. MATERIALS AND METHODS**

141 To better understand the interactions of stroke delimeter and grapple skidder and to test new processing
 142 techniques we create the stroke delimeter and grapple skidder agent based model (SDGS-ABM). The
 143 model was created using the agent based modeling tool NetLogo v5.0.5 (Wilensky). We present this
 144 model according to a modified version of the overview, design concepts, and details (ODD) protocol
 145 (Grimm, Berger, Bastiansen, et al.; Grimm, Berger, DeAngelis, et al.). The ODD protocol represents a
 146 well-adapted standard to communicate model descriptions consistently and effectively. The model has
 147 been developed in English units as the model is based on harvesting conditions in the Northeastern US
 148 and intended for the use in this region.

149 3.1. Purpose

150 The purpose of the model was to investigate the productivity of stroke delimeter and grapple skidder
 151 working on harvest tracts of different sizes and removal intensities. The goal was to gain knowledge
 152 about the productivity and time consumption of four different skidding and delimiting behaviors (Table
 153 2). As in real life, bunches consisted of different number of trees with different tree sizes. With results
 154 from this model it will be able to judge if the development of new technological communication features
 155 would in fact increase machine productivity and reduce waiting times.

156 Table 2: Description of four different skidding and delimiting behaviors as included in the model.

Scenario	Description
1	In this scenario there is no active communication between grapple skidder and stroke delimeter going on. There is no optimization of skidding times through additional processing information. This is our baseline scenario.

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- 2 In this scenario the stroke delimeter knows the processing time for each bunch. In addition to that the grapple skidder knows the traveling time for each bunch. Through the combination of the two sources of information the grapple skidder is able to select a bunch that will keep the waiting time for the stroke delimeter at a minimum.
 - 3 This scenario uses the same information as scenario 2, but in addition a process improvement feature for the stroke delimeter is introduced. The processing time for each tree is reduced by 1 second to improve the stroke delimeter productivity.
 - 4 This scenario is similar to scenario 1 but uses two grapple skidders instead of one. This will increase the stroke delimeter productivity by reducing the waiting time. (This scenario will not be analyzed in this paper as it is part of a more in-depth analysis)
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157

158 3.2. Entities, State Variables, and Scales

159 The model has four kinds of entities: grapple skidder, stroke delimeter, bunch, and square patches of
 160 land. Grapple skidder and stroke delimeter have no state variables, however, several pieces of
 161 information are recorded in global variables after each skidding cycle (Table 3). Each bunch has two
 162 state variables to describe the bunch size and the distance of the bunch to the landing following a
 163 previously laid out trail network. Patches are described by their patch size and the patch color based on
 164 landuse such as trail or forest land.

165 The grapple skidder is a moving agent that travels along a trail network and collects one bunch at a time.
 166 A bunch is placed along the trail network with a user-defined spacing between individual bunches.
 167 Bunches can only move when a grapple skidder picks them up and carries them to the landing where
 168 they are processed by the stroke delimeter. The stroke delimeter is a static agent that sits permanently at
 169 the landing and processes individual trees from a bunch. Several environmental variables are defined by
 170 the user; length of the main trail, removal per acre, bunch spacing, hardwood content, delimeter
 171 processing time improvement, stroke delimeter machine rate, and grapple skidder machine rate. The
 172 user can further choose to create bunches with equal sizes and select one of the four scenarios. Total
 173 width of the harvest tract is predefined at 612 feet. The road is 36 feet wide and next to a landing of 144
 174 feet by 300 feet. The trail system consists of one main trail with side trails leaving the main trail in a 45
 175 degree pattern and 60 foot trail spacing. The temporal extent of the model is the time it takes to skid
 176 and process all bunches along the trail.

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180 Table 3: State and global variables of the four model entities.

Entity	State / Global Variable
Grapple Skidder	Total waiting time in minutes Current waiting in minutes Current skidding time in minutes

	Total number of bunches skidded
Stroke Delimber	Total waiting time in minutes Current waiting time in minutes Current delimiting time in minutes Total number of bunches delimited
Bunch	Bunch size in tons Distance to the landing along trail in feet
Patch	Patch size (12 feet x 12 feet) Landuse type (trail, forest, landing, road)

181

182 **3.3. Process Overview and Scheduling**

183 During the model setup the following information is calculated and displayed in output monitors: area
184 harvested (acres), average bunch size (tons), length of main trail (feet), maximum skidding distance
185 (feet). The trail system is put in place during the model setup by the submodel “create trail” and
186 populated with bunches by the submodel “place bunches”. The model further includes the following
187 processes that are executed in this order during each time step.

188 *Skid bunch.* The grapple skidder moves to the nearest bunch along the main trail and brings the bunch
189 back to the landing. Once the main trail is cleared the skidder moves to the nearest bunch amongst the
190 side trails. If the user selects scenario 2 or 3 the skidder moves to the farthest bunch that is within the
191 distance that the skidder can travel during the time the delimber takes to process the previous bunch.
192 The total skidding time is calculated using a regional grapple skidder cycle time function (Table 4).

193 *Skid Two Bunches.* This process is only called for in scenario 4 when two grapple skidders are skidding
194 wood from the harvest tract. The process is similar to “skid bunch”, however, each skidder delivers
195 wood to their own drop zone at the landing, so that the stroke delimber has two bunches to work with.

196 *Process bunch.* During the first run of the “skid bunch” process the stroke delimber has no trees to
197 process and therefore has to wait for the skidder to come back. After the skidder brings a bunch the
198 sub-model “select trees” calculates the number of trees and individual tree sizes for the bunch. The
199 stroke delimber then processes one tree at a time. The time consumption for each tree is calculated
200 using a regional cycle time function for stroke delimber estimated from empirical data (Table 4).

201 *Update output.* This process updates all output monitors and advances time accordingly. Output
202 monitors record the following information: total skidded volume, total time consumption, current
203 grapple skidder time for this cycle, current grapple skidder waiting time for this cycle, total grapple
204 skidder waiting time, grapple skidder waiting time in percent of total time, current stroke delimber time
205 for this cycle, current stroke delimber waiting time for this cycle, total stroke delimber waiting time,
206 stroke delimber waiting time in percent of total time, system productivity (tons/PMH), grapple skidder
207 and stroke delimber total cost (\$), grapple skidder and stroke delimber harvest cost (\$/ac), and grapple
208 skidder and stroke delimber unit cost of production (\$/ton).

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214 Table 4: Description of values used in this model, including source of information.

Description	Value	Source
Cycle time equation for grapple skidder	Cycle Time (min) = $\exp(1.618 + 0.0005 \times \text{OneWayDistance (ft)})$	Hiesl and Benjamin (2013c)
Cycle time equation for stroke delimeter	Cycle Time (min) = $\exp(-1.247 + 0.099 \times \text{DBH (in)} - 0.135 \times \text{SpeciesGroup (1 = softwood, 2 = hardwood)})$	Hiesl and Benjamin (2013c)
Standard deviation for the spread of bunch sizes across a harvest site	0.8	unpublished results of Hiesl (2013)
Lamda-value for a poisson distribution that represents the distribution of tree diameters in a given bunch	8.43	unpublished results of Hiesl (2013)
Average removal intensity used in this study (in tons/acre)	40	unpublished results of Hiesl (2013)
Average spacing between individual bunches (in ft)	48	unpublished results of Hiesl (2013)

215

216 3.4. Design Concepts

217 The *basic principle* is to simulate the interactions between grapple skidder and stroke delimeter in four
 218 different scenarios that include (1) a “normal” harvest, (2) a harvest with perfect knowledge of
 219 processing times, (3) a harvest with perfect knowledge of processing times and increased delimiting
 220 speed, and (4) the use of two grapple skidders. The choice of perfect knowledge is based on potential
 221 technological developments (such as enhanced communication and location-tracking technology) within
 222 equipment cabs to accurately estimate processing times.

223 Grapple skidders *interact* with bunches by removing them from the trail and skidding them to the
 224 landing. The stroke delimeter processes one bunch, tree by tree, and in scenarios 2 and 3 estimates the
 225 processing time for each bunch. In these scenarios grapple skidder and stroke delimeter interact directly
 226 with each other by exchanging information which alters the skidding behavior of the grapple skidder.

227 The bunch size is *randomly* chosen using the average bunch size - calculated based on the user chosen
 228 removal intensity, bunch spacing, and length of trails – and a previously observed standard deviation of
 229 common bunch sizes (Table 4). The tree diameters in each bunch were randomly chosen using a
 230 previously observed Poisson distribution (Table 4). A differentiation is made between hardwoods and
 231 softwoods, as tree heights and volumes at a given diameter are different.

232

233 3.5. Initialization

234 The simulated harvest tract is created with a fixed height of 51 pixels (612 feet) and a user defined
235 length of between 40 and 210 pixels (480 to 2520 feet). The main trail is located in the center of the
236 world at a height of 26 pixels. Side trails are spurring in 45 degree angles at a spacing of 5 pixels (60
237 feet). Bunches are placed along the trail system with user defined bunch spacings. One grapple skidder
238 and one stroke delimeter are created in scenarios 1 to 3, while a second grapple skidder is created in
239 scenario 4. All equipment starts at the landing.

240 All time and productivity counters are set to zero. Each bunch has a randomly assigned bunch volume
241 and a distance to the landing calculated based on their location on the trail. The harvest area in acres is
242 calculated during the setup based on the trail system and a 25 foot swath on each side of the trail to
243 represent feller-buncher reach. The average bunch size, main trail length, and maximum skidding
244 distance are also calculated during the setup.

245

246 3.6. Submodels

247 *Create Trail:* The main trail is placed at the center of the World at a height of 26 pixels (312 feet).
248 Starting at the landing the side trails are spurring off in a 45 degree angle and reach all the way to the
249 boundary. The Spacing between trails is 5 pixels (60 feet).

250 *Place Bunches:* The number of bunches on the main trail and for each side trail are calculated during the
251 trail setup based on the user defined bunch spacing. Based on harvest area, removal intensity, and
252 number of bunches the average bunch size is calculated. Using a standard deviation of 0.8 (Table 4) the
253 bunch size for each bunch is randomly drawn. All bunches are placed at the end of each side trail and
254 trails are then populated with bunches towards the main trail. This feature represents common
255 harvesting techniques used in whole-tree harvesting in Maine. During the placement the model
256 periodically checks the total bunch size of all bunches placed on trails and compares it with the total
257 removal for the harvest tract and makes the necessary adjustments in bunch size if the total bunch size
258 is too high.

259 *Select Trees:* When a bunch is being processed by the stroke delimeter the bunch size is divided into
260 softwood (SW) and hardwood (HW) size based on the hardwood content chosen by the user. SW and
261 HW tree diameters are chosen from a Poisson distribution (Table 4). Each diameter is associated with an
262 average tree size. Individual tree sizes were calculated using Honer’s equations (Honer). Balsam fir
263 (*Abies balsamea* (L.) Mill.) and red maple (*Acer rubrum* L.) were used as reference species for softwood
264 and hardwood, respectively. We used average tree heights from unpublished data of Hiesl (2013) for the
265 calculations (Table 5). The tree size of each tree is added up until the bunch size for SW and HW is
266 reached.

267

268

269 Table 5: Tree diameter, height, and volume for softwood and hardwood trees used in the model (as
270 observed during a study by Hiesl (2013)).

Dbh (in)	Tree height (ft)	Tree volume (tons)	
		Softwood	Hardwood

4	37	0.078	0.071
5	43	0.138	0.126
6	48	0.215	0.201
7	53	0.315	0.280
8	56	0.429	0.408
9	59	0.563	0.540
10	61	0.711	0.657
11	63	0.880	0.854
12	64	1.059	1.030
13	66	1.270	1.241
14	70	1.532	1.512
15	74	1.825	1.819
16	76	2.112	2.116
17	78	2.425	2.440
18	78	2.719	2.736
19	79	3.054	3.080
20	86	3.570	3.658

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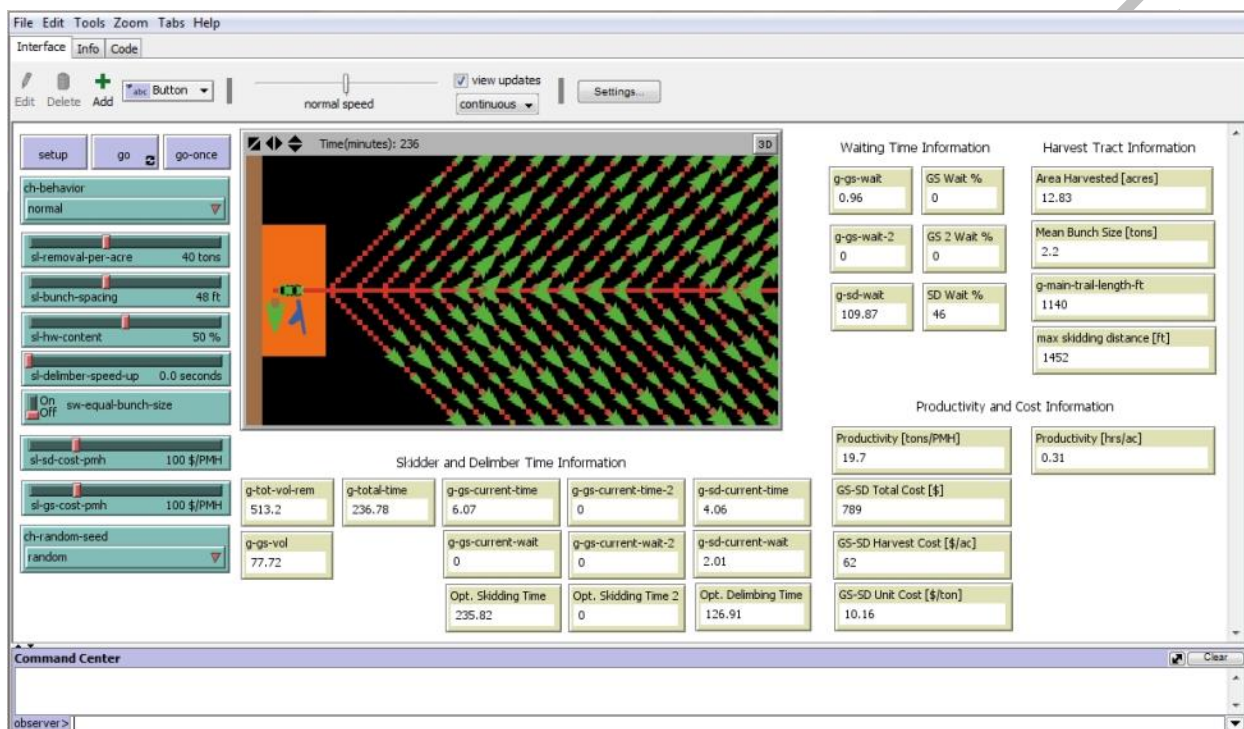
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273 3.7. Graphical User Interface

274 To increase the usability of this model a user friendly interface was created (Figure 3). This interface
275 includes several sliders with pre-defined settings to adjust various input variables such as the removal
276 per acre, bunch spacing, or hardwood content. Four grouping of output monitors exist to show the user
277 (1) important time information during each skid, (2) cumulative waiting time information, (3) cumulative
278 productivity and cost information, and (4) general harvest tract information.

279



280

281 Figure 3: Screenshot of the model interface.

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283 3.8. Simulation Analysis Methods

284 To analyze the effect of hardwood component on stroke delimeter and grapple skidder idle time the
 285 model was run using an average bunch spacing of 48 ft (Table 4), an average removal intensity of 40
 286 tons/acre (Table 4), and varying degrees of hardwood composition (Table 6). To analyze the effect of the
 287 different one skidder scenarios we run this setup for Scenarios 1 to 3. We used NetLogo’s BehaviorSpace
 288 module to run each configuration. The output was analyzed using the statistical software package R (R
 289 Core Team). To analyze the effect of bunch spacing and removal intensity on the baseline scenario
 290 (Scenario 1) we included a total of six bunch spacings and six removal intensities in the simulation.

291
 292 Table 6: Variables used in the simulation to analyze the effect of hardwood component on stroke
 293 delimeter and grapple skidder idle time.

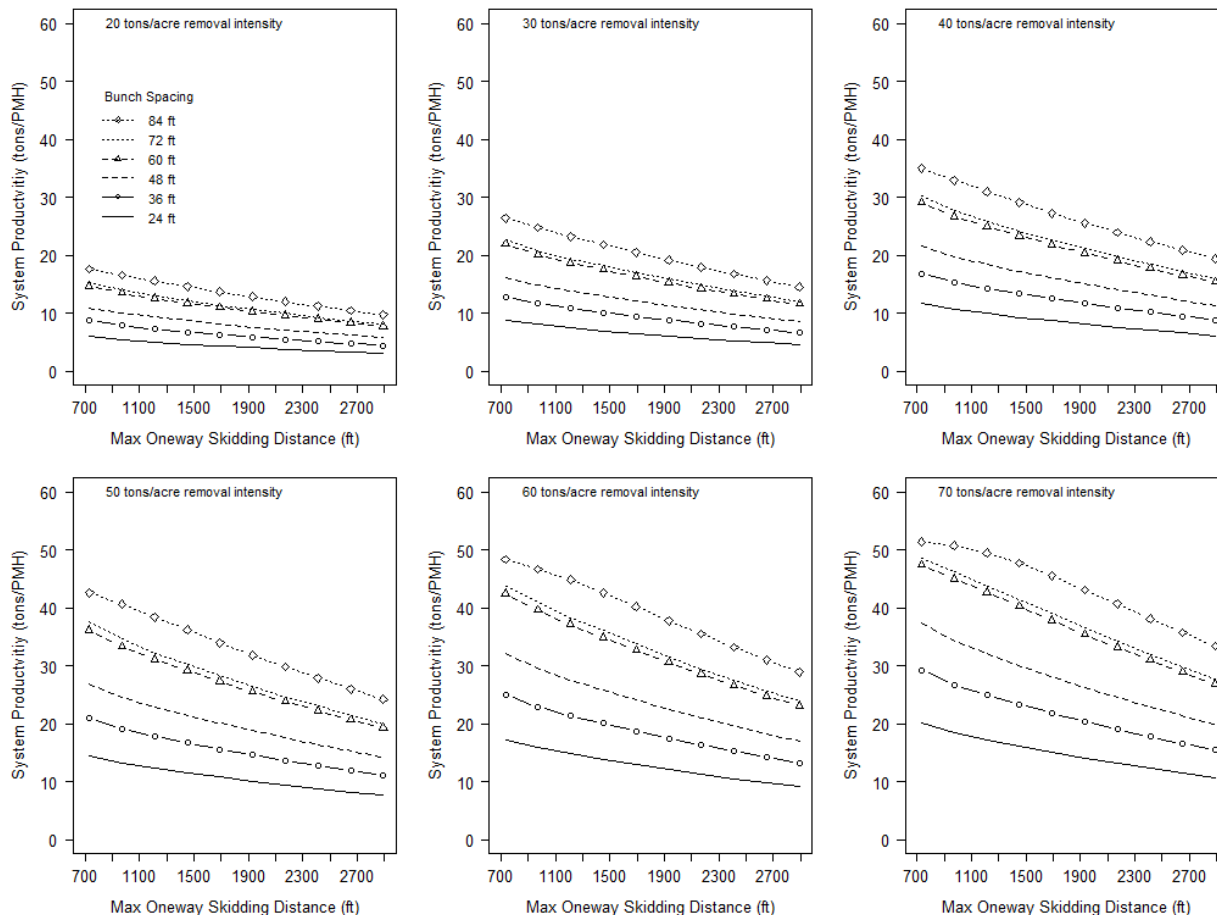
User-Defined Variable	Min-Value	Max-Value	Step-Size	# values tested
Scenario	1	3	1	3
Removal per acre (tons)	40	40	0	1
Bunch spacing (ft)	48	48	0	1
Hardwood content (%)	0	100	25	5
Max One-Way Skidding distance (ft)	732	2,892	240	10
			Parameter Combinations	150
			Simulations	15,000

294
 295 3.9. Sensitivity Analysis
 296 A local sensitivity analysis was conducted based on the Railsback and Grimm (2012) analysis structure.
 297 The goal of any sensitivity analysis is to understand how sensitive a model is to small changes in the
 298 value of input variables. Such information can help to validate the model structure by assessing whether
 299 or not specific sensitivities are represented in the model. A local sensitivity changes one input parameter
 300 at a time and therefore represents the sensitivity to such one parameter at a baseline of the other input
 301 parameter only. In contrast to that, a global sensitivity analysis changes several input values at the same
 302 time over a wide range of baseline scenarios to fully investigate the sensitivity of a model. For this local
 303 sensitivity analysis we increased the input values of three variables by 10% to calculate the sensitivity
 304 value. The baseline values of the three input variables reflect average skidding and delimiting conditions
 305 in Maine. These baseline values were determined from unpublished data of Hiesl (2013). The three input
 306 variables were chosen based on their known influence on system productivity from other research.

307
 308 **4. SIMULATION RESULTS**
 309

310 Results of baseline scenario (Scenario 1) of the model showed that the system productivity of grapple
 311 skidder and stroke delimeter was heavily influenced by skidding distance, removal per acre and bunch
 312 spacing (Figure 4). System productivity increased with increasing removal intensity, increasing bunch
 313 spacing and decreasing skidding distance. Bunch spacing ($p < 0.001$) and removal intensity ($p < 0.001$)
 314 clearly indicated a difference in system productivity.

315



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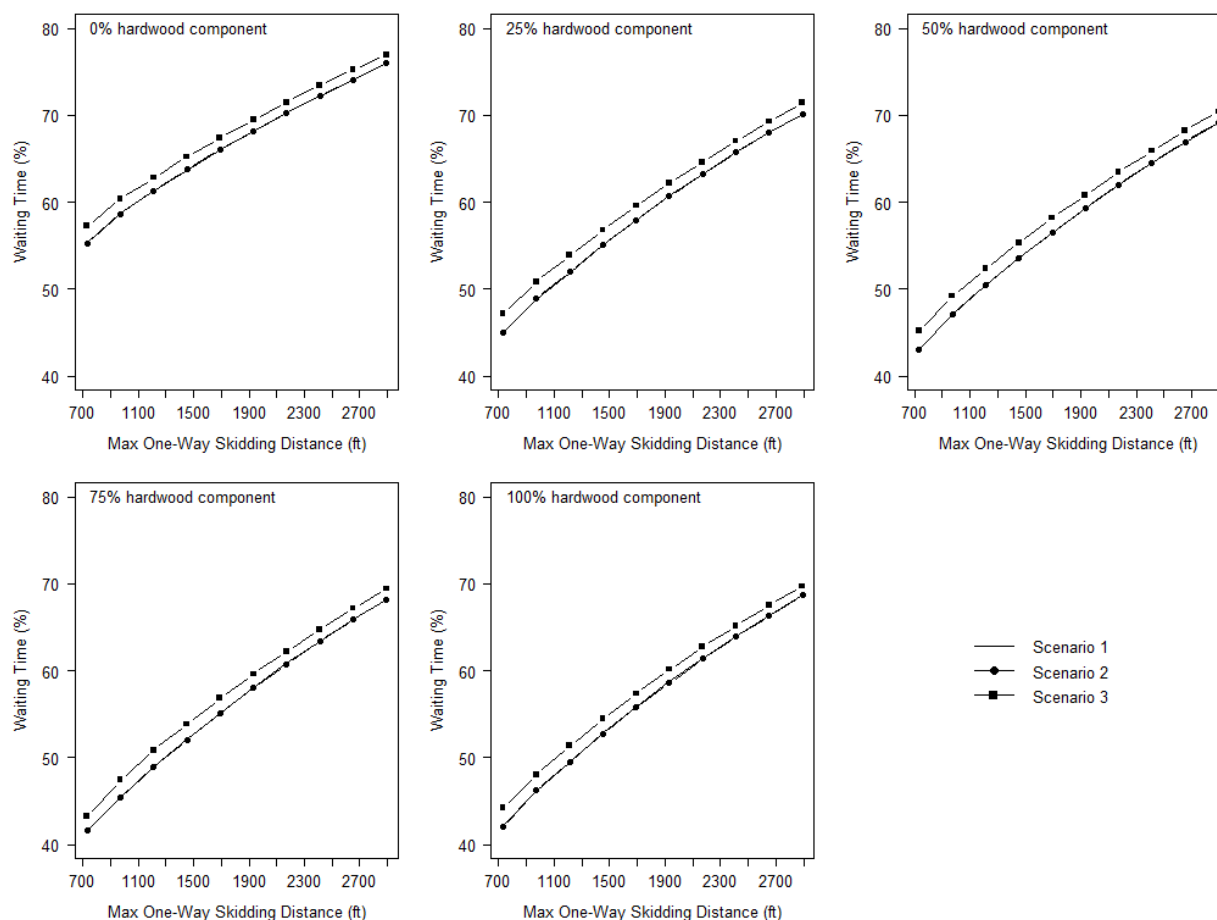
317 Figure 4: Grapple skidder and stroke delimeter system productivity based on removal intensity and
 318 bunch spacing with a 50% hardwood component when using the baseline scenario (Scenario 1). PMH =
 319 productive machine hours.

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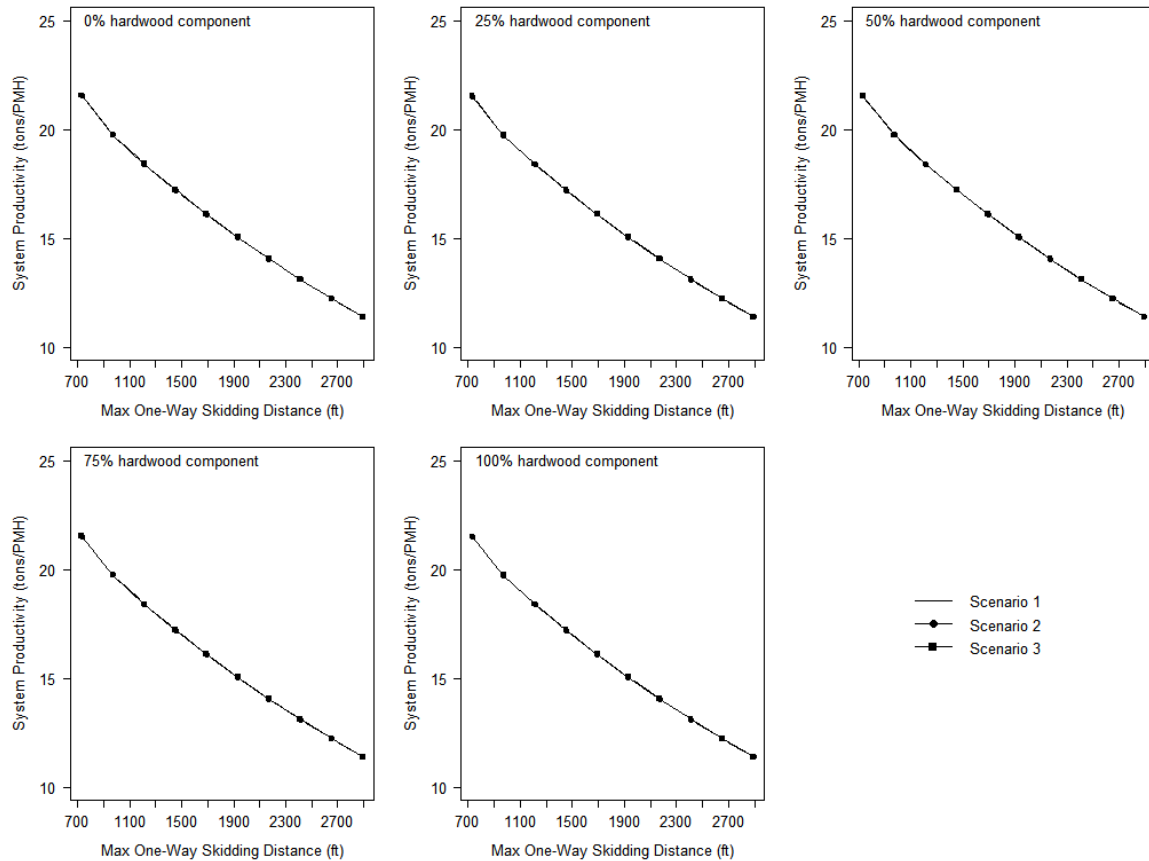
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323 The analysis of the effect of hardwood component on stroke delimeter waiting time showed that there is
 324 a reduction in waiting time with increasing hardwood component ($p < 0.001$). This reduction is up to 13%
 325 at short skidding distances and decreases to 7% at the longest skidding distance (Figure 5). No difference
 326 was found in the stroke delimeter waiting time between Scenario 1 and Scenario 2 ($p = 0.999$), however,
 327 there was a difference between Scenario 3 and the other two scenarios ($p = 0.004$). Grapple skidder
 328 waiting time was not affected by the change in hardwood component and stayed below 1%.



329
 330 Figure 5: Stroke delimeter waiting time based on various hardwood components and an average bunch
 331 spacing of 48 ft and a removal intensity of 40 tons per acre. No difference was found between Scenario
 332 1 and Scenario 2 and both lines are approximately on top of each other.

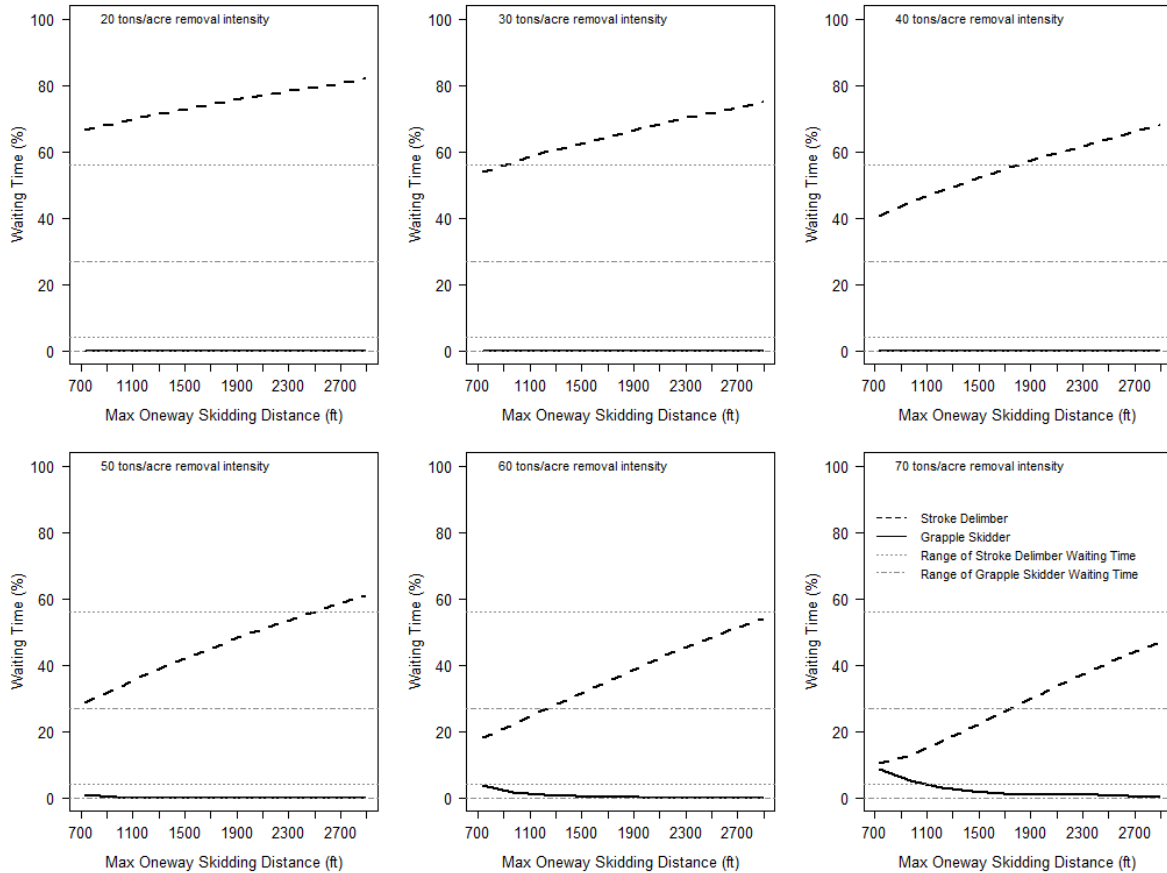
333 Even though there is a decrease in stroke delimeter idle time with increasing hardwood component, our
 334 results show that the system productivity is not affected by hardwood component ($p = 0.922$). Further,
 335 there was no difference ($p = 0.998$) found in system productivity between the three tested scenarios
 336 (Figure 6). Thus, the productivity stays the same whether or not the hardwood component increases,
 337 new GIS/GPS tools are used (Scenario 2), or the stroke delimeter increases processing speed (Scenario 3).
 338 The only influential factor on system productivity is skidding distance ($p < 0.001$). An increase in skidding
 339 distance causes a decrease in system productivity.



340

341 Figure 6: System productivity of a grapple skidder and stroke delimeter system based on various
 342 hardwood components and an average bunch spacing of 48 ft and a removal intensity of 40 tons per
 343 acre . The productivity of all three scenarios is similar and thus the individual lines are overlaying each
 344 other.

345 Waiting time data from Figure 1 shows that a stroke delimeter generally waits between 4% and 56% of
 346 the time, while a grapple skidder waits between 0% and 27%. These values have been collected from
 347 harvest sites with removal intensities ranging from 25 tons/acre up to 67 tons/acre. The waiting times
 348 produced by this model (Figure 7) are similar to the range of observed waiting times. This shows that the
 349 model is an accurate representation of a stroke delimeter and grapple skidder harvesting system.



350

351 Figure 7: Waiting time for grapple skidder and stroke delimeter based on skidding distance and removal
 352 intensity with a 50% hardwood component using the baseline scenario (Scenario 1).

353

354 5. MODEL EVALUATION

355 The relationship between system productivity and site specific variables such as skidding distance and
 356 removal intensity, in combination with the correct representation of waiting times leads us the
 357 assumption that this model is valid and well calibrated. To increase the usefulness of this model to other
 358 researchers and the logging community, however, it is crucial to test the model for its sensitivity to
 359 parameter combinations.

360 Local and global sensitivity analyses were used to evaluate the sensitivity of our model to a change in
 361 input variables. Results showed that average bunch size had the greatest impact on system productivity,
 362 followed by skidding distance (Table 7). The impact of hardwood content on system productivity was
 363 very low compared to the other two input variables. Such an analysis is a snapshot of the effect of input
 364 variables on system productivity based on baseline conditions that represent average harvesting
 365 conditions in Maine. To gain more insight of the effect of these variables based on a variety of
 366 harvesting conditions we conducted a global sensitivity analysis (Figure 8).

367 The global sensitivity analysis shows that the effect of bunch size on system productivity is less intensive
 368 at short skidding distances and high bunch sizes and increases with skidding distance and a reduction in

369 bunch size. The effect of skidding distance on system productivity intensifies with an increase in skidding
370 distance. This effect, however, is reduced with an increase in bunch size. A higher softwood component
371 increases system productivity at short skidding distances and high bunch sizes but loses intensity with
372 longer skidding distances.

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375

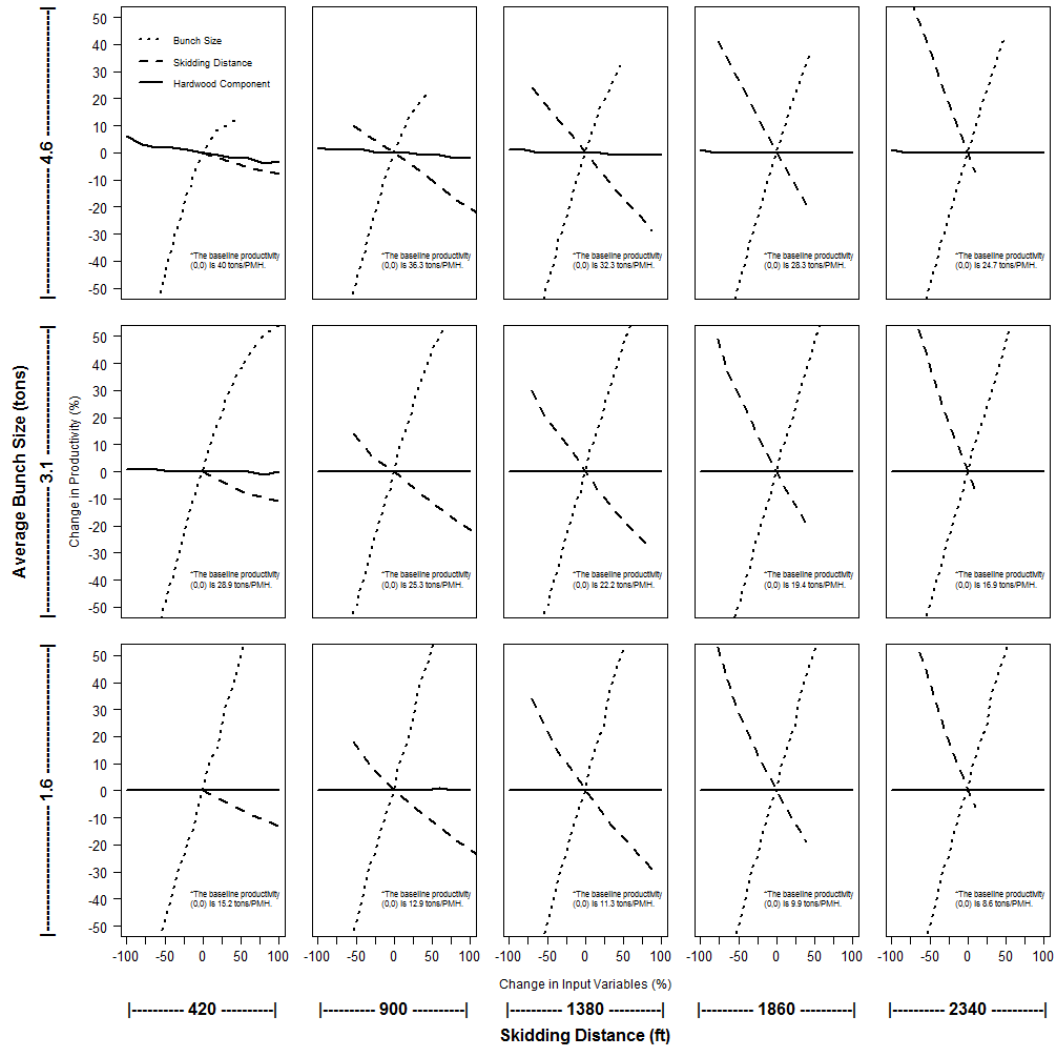
376 Table 7: Local Sensitivity Analysis of three input variables.

Parameter	Reference value	Sensitivity value	Change in productivity (%)	Change in productivity (tons/PMH)
Skidding Distance (ft)	1,380	-10.51	-4.82	-1.05
Hardwood Content (%)	50	-0.20	-0.09	-0.02
Average Bunch Size (tons)	3.0	22.44	10.29	2.24

377 **Note:** Parameter values were increased by 10%.

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380 Figure 8: Global Sensitivity Analysis of three input variables based on baseline conditions consisting of a
 381 variety of skidding distances (x-axis), average bunch sizes (y-axis), and a 50% hardwood component.

382 6. DISCUSSION

383 The use of agent based modeling in forestry is fairly new. Agent-based modeling has been used to
 384 investigate and model harvesting decision making of landowners (Leahy et al.), simulate landscape-scale
 385 forest ecosystem dynamics (Seidl et al.), and model harvesting scenarios in mangrove forest plantations
 386 (Fontalvo-Herazo et al.). Our model is one of the first to apply an agent-based approach to production
 387 forestry in a developed country. We further created a graphical user interface in NetLogo (Wilensky)
 388 that allows users to vary input variables such as removal intensity, hardwood content, and bunch
 389 spacing.

390 This model uses cycle time equations specifically developed for harvesting systems in Maine. In addition
 391 to that all the values and probabilities used in this simulation are from unpublished data of a harvesting
 392 cycle time and productivity study by Hiesl (2013). Such empirical data increases the applicability and
 393 plausibility of this model. For example, the sensitivity analysis returned skidding distance and bunch size

394 as important factors affecting system productivity. Skidding distance is a well-known factor that affects
395 skidder productivity and has been reported by several researchers (Han et al.; Kluender et al.; Andersson
396 and Evans; Hiesl). Bunch size, or payload, has also been described as a factor influencing grapple skidder
397 productivity (Kluender et al.; Wang et al.). As our model aligns with this previous literature we are
398 confident that the core model dynamics are accurate and well calibrated including the relationship
399 between input variables and system productivity. With the open source characteristic of this model it is
400 possible for other researchers to extend the existing model to include other harvesting systems and
401 management treatments. Such extension could include further calibration, development of other sub-
402 models, or the inclusion of new data. The benefit of our agent-based model is that these changes are
403 relatively simple to implement.

404 Our results showed that an increase in hardwood component can reduce the waiting time for a stroke
405 delimeter but has no effect on the waiting time of a grapple skidder. This results is not surprising as the
406 literature indicates that the stroke delimeter processing time is higher for hardwood than it is for
407 softwood species (Hiesl). One reason for this increase in time consumption can be found in the larger
408 branch size and the increased number of forks in the crown. Research with harvesters showed that a
409 large branch size negatively affects processing speed and productivity (Glöde). A stroke delimeter uses a
410 similar movement to delimb trees as a harvester does, so it is a reasonable assumption that the same
411 reasons apply here. With the lowest waiting time being approximately 40% it is not surprising that the
412 grapple skidder waiting time is close to zero. Even though there is a negative effect of hardwoods on
413 processing speed, there was a positive effect on waiting time. This is due to the large number of excess
414 time that a stroke delimeter has before the grapple skidder can deliver a new bunch. The presented
415 simulation, however, was done based on average bunch spacing and removal intensity. In many
416 situations a land manager or logging contractor has to deviate from these standards and may encounter
417 a more positive or negative effect of a change in hardwood component.

418 A decrease in stroke delimeter idle time, however, does not necessarily mean that there will be an
419 increase in system productivity. Our results showed that hardwood component did not affect system
420 productivity. This can be attributed to the fact that in the presented case the skidding time is not
421 affected by the species mix in each bunch and thus stays the same regardless of hardwood component.
422 This further means that the overall time consumptions stays the same, even though the stroke delimeter
423 spends less time waiting for a new bunch. Research indicated that grapple skidder productivity is
424 affected by payload (Hiesl; Kluender et al.; Wang et al.; Y. Li et al.) and thus the results might be
425 different when changing the average bunch size in our simulation. In our analysis, however, we were
426 interested in the effect of varying hardwood components on system productivity and machine idle time
427 when operating under average harvesting conditions. The fact that there is no effect on system
428 productivity therefore indicates that mixed-wood and hardwood stands in Maine can be treated without
429 losing any productivity or increasing harvest costs.

430 When looking at the system productivity the results further showed that there is no difference in
431 productivity between the three tested scenarios. The surprise was that the use of GIS/GPS (Scenario 2)
432 did not result in any production increase. One reason for this might be the use of one main trail only.
433 This fact limits the grapple skidder in the number of bunches that can be chosen to minimize stroke
434 delimeter waiting time. Another reason might be the chosen behavior rule of selecting the bunch that is
435 farthest away but does not cause any more stroke delimeter delay. This behavior rule did not include the
436 clearing of the main trail first and thus limited the number of bunches that were accessible. The third
437 scenario included an increase in processing speed of 1 second per tree. This increase in processing time
438 resulted in an increase in stroke delimeter idle time. This can be attributed to the fact that the grapple

439 skidder was not delivering bunches any faster and thus the increased processing time left more time for
440 the stroke delimeter to wait for the grapple skidder.

441 In Figure 4 system productivity is shown for varying removal intensities and bunch spacings. Individual
442 productivity curves are fairly uniformly distributed among the different bunch spacings with the
443 exception of the 60 and 72 ft bunch spacing. These two productivity curves are very close and almost
444 overlay each other. The reason for this lies in the bunch placing process of the simulation. Bunches were
445 placed on a side trail starting at the end of a trail and then spacing them by the user defined bunch
446 spacing. For all bunch spacings the number of bunches per side trail decreased with increasing bunch
447 spacing, with the exception of the 60 and 72 ft bunch spacing. In this special case, the number of
448 bunches in each side trail stayed the same, with the exception of a few side trails at the end of the
449 harvest block. Lengthening or shortening the side trails only shifted this process to a different pair of
450 bunch spacings. It is important to notice, however, that this effect also happens at real harvest sites, and
451 thus a change in bunch spacing might not have the sought after effect of increasing bunch size.

452 Next, this model needs to be applied to investigate system productivity change, and skidder and
453 delimeter wait time that emerge from real world harvesting scenarios. For instance, an analysis might
454 seek to answer the question whether or not an investment in various types of communication or spatial
455 awareness technology will result in any productivity gains across varying stand and site conditions, and if
456 so, whether or not this investment will pay for itself during the lifetime of these machines. Further
457 economic calculations should include the unit cost of production as a measure of applicability of any
458 system in the real world at the current market conditions. Many additional alternative management
459 configurations are also possible with an agent-based system because the design of machine behavior
460 and machine-machine interaction is greatly simplified over traditional approaches.

461 **7. CONCLUSION**

462 Our conclusion is that under average harvesting conditions in Maine it does not pay to invest in new
463 GIS/GPS technology, at least not with the proposed behavior rules for such a system. Further, an
464 increased hardwood component, under these average harvesting conditions, does not affect system
465 productivity. This leaves current market conditions as one of the few limitations of treating mixed-wood
466 and hardwood stands in Maine.

467 Unless the system of grapple skidder and stroke delimeter is de-coupled, logging contractors and land
468 managers have to accept that under average harvesting conditions the stroke delimeter will wait for
469 trees to be processed at least 40% of its operational time. With machine rates upwards of \$100
470 USD/PMH this means that over \$40 USD/PMH are paid for sitting at the landing and waiting for wood.
471 This is money spent without getting any return. Clearly there is a need to find new ways to use these to
472 machines to further reduce the waiting time of either machine and to limit to money spent on processes
473 that do not return any revenue.

474

475 **8. LITERATURE CITED**

476 Adebayo, A. B., H.-S. Han, and L. Johnson. Productivity and cost of cut-to-length and whole-tree
477 harvesting in a mixed-conifer stand. *Forest Products Journal* 57 (6): 59–69.

- 478 Andersson, B., and C. M. Evans. *Harvesting overmature aspen stands in central Alberta. Special Report*
479 *SR-112*. Pointe Claire, PQ, Canada: Forest Engineering Research Institute of Canada (FERIC).
- 480 Arij, K., J. P. Caspersen, T. A. Jones, and S. C. Thomas. A selection harvesting algorithm for use in spatially
481 explicit individual-based forest simulation models. *Ecological Modelling* 211 (3-4): 251–266.
- 482 Baumgras, J. E., C. C. Hassler, and C. B. LeDoux. Estimating and validating harvesting system production
483 through computer simulation. *Forest Products Journal* 43 (11/12): 65–71.
- 484 Bazghandi, A. Techniques , Advantages and Problems of Agent Based Modeling for Traffic Simulation.
485 *International Journal of Computer Science Issues* 9 (1): 115–119.
- 486 Bolding, M. C. Increasing Forestry Machine Utilization. *Sawmill & Woodlot* February: 23–27.
- 487 Borshchev, A., and A. Filippov. From System Dynamics and Discrete Event to Practical Agent Based
488 Modeling: Reasons, Techniques, Tools. In *The 22nd International Conferences of the System Dynamics*
489 *Society*, 23. Oxford, England.
- 490 Bradley, D. P., F. E. Biltonen, and S. A. Winsauer. *A Computer Simulation of Full-Tree Field Chipping and*
491 *Trucking. Research Paper NC-129*. St. Paul, MN, USA: U.S. Department of Agriculture Forest Service,
492 North Central Forest Experiment Station.
- 493 Dixon, G. E. *Essential FVS: A user's guide to the Forest Vegetation Simulator. Internal Report (Revised:*
494 *October 16, 2012)*. Fort Collins, CO: U.S. Department of Agriculture Forest Service, Forest Management
495 Service Center.
- 496 Epstein, J. M. Agent-Based Computational Models And Generative Social Science. *Complexity* 4 (5): 41–
497 60.
- 498 Fontalvo-Herazo, M. L., C. Piou, J. Vogt, U. Saint-Paul, and U. Berger. Simulating harvesting scenarios
499 towards the sustainable use of mangrove forest plantations. *Wetlands Ecology and Management* 19 (5):
500 397–407.
- 501 Garner, G. J. *Simulation: A decision making aid for managers. Technical Report TR-30*. Pointe Claire, PQ,
502 Canada: Forest Engineering Research Institute of Canada (FERIC).
- 503 Gilbert, N. *AGENT-BASED MODELS. Quantitative Applications in the Social Sciences*. Thousand Oaks, CA,
504 USA: Sage Publications, Inc.
- 505 Glöde, D. Single- and double-grip harvesters - productive measurements in final cutting of shelterwood.
506 *International Journal of Forest Engineering* 10 (2): 63–74.
- 507 Goulet, D. V, R. H. Iff, and D. L. Sirois. Tree-to-Mill Forest Harvesting Simulation Models: Where Are We?
508 *Forest Products Journal* 29 (10): 50–55.

- 509 Grimm, V., U. Berger, F. Bastiansen, S. Eliassen, V. Ginot, J. Giske, J. Goss-Custard, et al. A standard
510 protocol for describing individual-based and agent-based models. *Ecological Modelling* 198 (1-2): 115–
511 126.
- 512 Grimm, V., U. Berger, D. L. DeAngelis, J. G. Polhill, J. Giske, and S. F. Railsback. The ODD protocol: A
513 review and first update. *Ecological Modelling* 221 (23): 2760–2768.
- 514 Han, H.-S., H. W. Lee, and L. R. Johnson. Economic feasibility of an integrated harvesting system for
515 small-diameter trees in southwest Idaho. *Forest Products Journal* 54 (2): 21–27.
- 516 Hiesl, P. Productivity standards for whole-tree and cut-to-length harvesting systems in Maine. Master
517 Thesis, Orono, ME, USA: University of Maine - School of Forest Resources.
- 518 Hiesl, P., and J. G. Benjamin. A multi-stem feller-buncher cycle-time model for partial harvest of small
519 diameter wood stands. *International Journal of Forest Engineering* 24 (2): 101–108.
- 520 Hiesl, P., and J. G. Benjamin. *Harvesting Equipment Cycle Time and Productivity Guide for Logging*
521 *Operations in Maine*. Miscellaneous Publications 762. *Miscellaneous Publications 762*. Orono, ME, USA:
522 Maine Agricultural and Forest Experiment Station.
- 523 Hiesl, P., and J. G. Benjamin. Applicability of international harvesting equipment productivity studies in
524 Maine, USA: A literature review. *Forests* 4 (4): 898–921.
- 525 Hiesl, P., and J. G. Benjamin. Estimating processing times of harvesters in thinning operations in Maine.
526 *Forest Products Journal*.
- 527 Holzleitner, F., K. Stampfer, and R. Visser. Utilization rates and cost factors in timber harvesting based on
528 long-term machine data. *Croatian Journal of Forest Engineering* 32 (2): 501–508.
- 529 Honer, T. G. *Standard volume tables and merchantable conversion factors for the commercial tree*
530 *species of central and eastern Canada*. Information Report FMR-X-5. Ottawa, Ontario: Canadian
531 Department of Forestry and Rural Development. Forest Management Research and Service Institution.
- 532 Huth, A., M. Drechsler, and P. Köhler. Multicriteria evaluation of simulated logging scenarios in a tropical
533 rain forest. *Journal of environmental management* 71 (4): 321–33.
- 534 Kluender, R., D. Lortz, W. McCoy, B. J. Stokes, and J. Klepac. Productivity of Rubber-tired Skidders in
535 Southern Pine Forests. *Forest Products Journal* 47 (11/12): 53–58.
- 536 Leahy, J. E., E. G. Reeves, K. P. Bell, C. L. Straub, and J. S. Wilson. Agent-Based Modeling of Harvest
537 Decisions by Small Scale Forest Landowners in Maine, USA. *International Journal of Forestry Research*
538 2013: 1–12.
- 539 Leon, B., and J. G. Benjamin. *A Survey of Business Attributes, Harvest Capacity and Equipment*
540 *Infrastructure of Logging Businesses in the Northern Forest*. *The Northern Forest Logging Industry*
541 *Assessment*. Orono, ME, USA: University of Maine.

542 Li, R., A. Weiskittel, A. R. Dick, J. A. K. Jr, and R. S. Seymour. Regional stem taper equations for eleven
543 conifer species in the Acadian Region of North America: Development and Assessment. *Northern Journal*
544 *of Applied Forestry* 29 (1): 5–14.

545 Li, Y., J. Wang, G. Miller, and J. McNeel. Production economics of harvesting small diameter hardwood
546 stands in central Appalachia. *Forest Products Journal* 56 (3): 81–86.

547 Manson, S. M. Validation and Verification of Multi-Agent Systems. In *Complexity and Ecosystem*
548 *Management: The Theory and Practice of Multi-Agent Approaches*, ed. M. A. Janssen, 63–74.
549 Northhampton, MA,USA: Edward Elgar Publishers.

550 McCaskill, G. L., W. H. McWilliams, C. J. Barnett, B. J. Butler, M. A. Hatfield, C. M. Kurtz, R. S. Morin, W. K.
551 Moser, C. H. Perry, and C. W. Woodall. *Maine's Forests 2008. Resource Bulletin NRS-48*. Newtown
552 Square, PA: U.S. Department of Agriculture Forest Service, Northern Research Station.

553 Polley, H. Anwendung eines digitalen stochastischen Modells zur Kalkulation technologisch bedingter
554 Wartezeiten. *Beiträge für die Forstwirtschaft* 21 (2): 72–77.

555 Purfürst, F. T. Learning Curves of Harvester Operators. *Croatian Journal of Forest Engineering* 31 (2): 89–
556 97.

557 R Core Team. *R: A language and environment for statistical computing*. Vienna, Austria: R Foundation for
558 Statistical Computing.

559 Railsback, S. F., and V. Grimm. Sensitivity, Uncertainty, and Robustness Analysis. In *Agent-Based and*
560 *Individual-Based Modeling. A Practical Introduction*, 291–308. Princeton, NJ, USA: Princeton University
561 Press.

562 Seidl, R., W. Rammer, R. M. Scheller, and T. A. Spies. An individual-based process model to simulate
563 landscape-scale forest ecosystem dynamics. *Ecological Modelling* 231: 87–100.

564 Spinelli, R., B. R. Hartsough, and N. Magagnotti. Productivity standards for harvesters and processors in
565 Italy. *Forest Products Journal* 60 (3): 226–235.

566 Wang, J., and C. B. LeDoux. Estimating and Validating Ground-Based Timber Harvesting Production
567 Through Computer Simulation. *Forest Science* 49 (1): 64–76.

568 Wang, J., C. Long, and J. McNeel. Production and cost analysis of a feller-buncher and grapple skidder in
569 central Appalachian hardwood forests. *Forest Products Journal* 54 (12): 159–167.

570 Weiskittel, A., and R. Li. *Development of regional taper and volume equations: Hardwood species*. Ed. B.
571 E. Roth. *Cooperative Forestry Research Unit 2011 Annual Report*. Orono, ME, USA: University of Maine.

572 Wilensky, U. *NetLogo*. Evanston, IL, USA: Center for Connected Learning and Computer-Based Modeling,
573 Northwestern University.

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