

1 **Revised Report: EMC-2017-007**

2 John Battles 8/5/2023

3 **The Life Cycle of Standing Dead Trees: Implications for Forest Management in the Sierra Nevada.**

4 ...standing dead trees may result from a number of agencies, such as fire, bark beetles,
5 tree diseases, flooding and drought. Once produced, they become of concern to
6 foresters...

7 Keen 1929

8 INTRODUCTION

9 **Relevance to management.** Standing dead trees (i.e., snags) are vital but ephemeral elements of
10 the forest. They represent the transition from living trees where entropy is actively delayed by the input
11 of energy to downed wood where the direct contact with soil microbes speeds decay (Franklin et al.
12 1987). While they remain standing, these trees provide essential habitat for wildlife; they store a
13 significant amount of carbon; and they present potential hazards (Hilger et al. 2012). Thus as Keen
14 (1929) noted, snags are a concern to foresters and to the practice of forestry.

15 At the heart of the long-standing conflict is the fact that snags provide benefits to wildlife.
16 Raphael and White (1984) documented 18 different species of cavity nesting birds using snags in just
17 one Sierra Nevada forest. The California Forest Practice Rules¹ estimate that snags contribute to the
18 habitat needs of 160 wildlife species [Technical Rule Addendum No.2: C.4.a]. Moreover large snags
19 serve as critical habitat (i.e., nest, den, or resting sites) for three forest species of conservation concern:
20 California spotted owl (*Strix occidentalis occidentalis*, Gutiérrez et al. 2017), Pacific fisher (*Martes*
21 *pennantia*, Weir et al. 2012), and the marten (*Martes americana*, Spencer et al. 1983). At the same time,
22 snags pose fire and safety risks (Knapp 2015). These very real threats justified the earlier requirement
23 (1947-1976) in the California Forest Practice Rules to remove nearly all large snags during timber
24 harvest operations (Raphael and White 1978). Today the Forest Practice Rules clearly recognize the
25 value of snags. For example, in the Appendix Technical Rule Addendum No. 2, snags are a “significant

26 factor” in the biological habitat condition of a timber harvest plan [C.4, p.45¹]. Snags are also
27 acknowledged as an important part of the vegetation structure along water courses that deserve
28 protection [916.4, 936.4, 956.4 Watercourse and Lake Protection (b).g.6¹]. Thus the Forest Practice
29 Rules stipulate the retention of snags in harvest plans [14 CCR § 919.1 (939.1, 959.1¹)]. However there
30 are multiple exceptions to the retention stipulation and there is no established practice for managing snag
31 density. Clearly, Californians expect foresters to consider snags and manage them to meet objectives of
32 both fire hazard reduction and wildlife habitat. Unfortunately, the basic information needed to make
33 decisions that influence snags is inadequate. *The primary goal of this research is to provide the*
34 *necessary scientific basis to inform snag retention guidelines.*

35 A more recent consideration for forest practice in California is the impact management has on the
36 carbon balance. Specifically timber harvests need to consider the capacity of the forest to sequester
37 carbon dioxide following proposed operations [Article 4, 4551 (b) (1)¹]. Given the recent increases in
38 tree mortality (van Mantgem and Stephenson 2007, Young et al. 2017), snags will play an increasingly
39 important role in the carbon dynamics of California forests – a role that is poorly quantified at present.
40 While current carbon impact assessments of timber harvest plans do account for carbon in snags, better
41 information on carbon dynamics in snags can make these assessments more accurate. *Thus the secondary*
42 *goal of this research is to improve our understanding of the contribution of snags to carbon storage in*
43 *the Sierran mixed conifer forest.*

44
45 **Objectives.** This research addresses critical monitoring questions identified in the Effectiveness
46 Monitoring Committee's (EMC) strategic plan. Snags are a major component of the EMC theme:
47 "Wildlife Habitat: Structures." The Forest Practice Rules [C4a, p. 45¹] specifically recommend that given

¹ California Forest Practice Rules 2022. https://bof.fire.ca.gov/media/y5rfw50b/2022-fpr-and-fpa_ada.pdf

48 the importance of snags as den and wildlife trees, “(t)he degree of snag recruitment over time may be
49 considered.” Yet the dynamics of snags are complex (Cousins et al. 2015). During the time a dead tree
50 remains standing, a typical sequence of changes occurs leading to an overall reduction in tree size. Tree
51 volume declines through loss of leaves, twigs, and branches, which fall to join the downed wood on the
52 forest floor. Concurrent with these dimensional reductions are changes to the tree’s physical and
53 chemical properties caused by weathering, decomposition, and insect activity. This complexity makes
54 managing snags difficult. For regulations that either require snag retention or that count snags for
55 stocking compliance, key questions are: What level of snag retention during operations is sufficient,
56 given how long they will last? Should active snag recruitment (i.e., girdling) be an option for snag
57 management or mitigation? What is the impact of snag retention and recruitment on carbon cycling?

58 Because of the decay process, the value of snags as resource for wildlife evolves with time since
59 death (Raphael and White 1984). In the early stages, snags are an important source of food for wood
60 boring insects, which in turn, are the primary food source for woodpeckers. With the decomposition of
61 wood, wildlife use value changes (Lorenz et al. 2015). As cavities form, they provide critical den and
62 nest sites. At the same time, the carbon stored in the snag decreases along with mechanical stability
63 (Cousins et al. 2015). Just as foresters do with live trees, they can plan for a sustainable presence of
64 snags if they have adequate demographic information. However, we have limited information on the
65 longevity of snags. The fall rates vary by species and tree size. For Sierran mixed conifer forests, annual
66 fall rates range from 7% yr⁻¹ to 14% yr⁻¹ (Battles et al. 2015). There is even less information on the rate
67 of decay. We know the steps, but we do not know how long it takes for a recently dead tree to become a
68 suitable wildlife habitat tree.

69 Finally as snags fall, they add to the downed wood in the forest. The decay rate of the downed
70 wood then determines the longevity of this additional surface fuel. Usually this downed wood is in
71 steady state with losses (decay) matching additions (snag fall). However in stands in the Sierra Nevada

72 with drought-related mortality, there is the potential for large pulses and sustained additions in downed
73 wood that in turn increase the fire hazard. A key gap in our knowledge is that we do not have good
74 measures of decay rates of downed wood in Sierran conifer forests that can guide management and
75 policy decisions relating to snag retention and recruitment.

76 To answer these questions and inform forest practice, we relied on a rare resource -- a long-term
77 snag inventory and monitoring study at Blodgett Forest Research Station. The inventory has been
78 repeated at irregular intervals: 1989, 1994, 1995, 2005, and 2012. For this project, we identified three
79 specific objectives:

80

81 1. Extend the snag inventory at Blodgett Forest to develop species and size specific snag
82 fall rates.

83

84 2. Implement a new monitoring protocol that tracks cohorts of new snags on an annual
85 basis to quantify the development of important habitat elements (e.g., nest cavities) and
86 changes in the carbon density of snags.

87

88 3. Establish a long-term study of downed wood decay rates.

89

90 METHODS

91 **Site description.** Blodgett Forest Research Station (BFRS) is a research forest (4,355 ac in area)
92 in the central Sierra Nevada, California, USA (38°52N; 120°40W). Six native tree species are commonly
93 found in mixtures of varying proportions: white fir (*Abies concolor*), incense cedar (*Calocedrus*
94 *decurrens*), coast Douglas-fir (*Pseudotsuga menziesii* var. *menziesii*), sugar pine (*Pinus lambertiana*),
95 ponderosa pine (*Pinus ponderosa*), and California black oak (*Quercus kelloggii*). The forest is

Table 1. Forest structure and composition in Compartment 160 at Blodgett Forest Research Station. The number in parentheses is the standard error (se) of the mean

	1983 mean (se)	2013 mean (se)
Density (trees ac ⁻¹)	247 (27)	166 (17)
Basal area (ft ² ac ⁻¹)	164.6 (17.0)	169.0 (15.2)
Relative Dominance (%)		
Black oak	18	9
Douglas-fir	12	21
Incense-cedar	15	18
Ponderosa pine	26	16
Sugar pine	10	12
White fir	18	25

97 representative of a productive mixed-conifer forest that occurs between 4,000 to 6,000 ft in elevation on
98 the western Sierra Nevada slope. BFRS occurs at the low end of this elevation gradient (3,900 to 4,800
99 ft).

100 The regional climate is Mediterranean, characterized by a summer drought period and mild
101 winters. Most precipitation occurs during the winter and spring, averaging 65 in annually. Average daily
102 summer temperatures range between 57° and 81° F while average winter temperatures are between 16°
103 and 32° F (BFRS data, <https://forests.berkeley.edu/forests/blodgett>). The nutrient-rich soils at BFRS are
104 formed from andesitic parent materials (Heald and Barrett 1999).

105 Snag monitoring at BFRS was conducted in Compartment 160, a 59-ac management unit. The
106 aspect is predominantly east-facing, with slopes variable from 5-18% and elevations ranging from to
107 4,000 to 4,400 ft. The site was extensively cut and likely burned in 1913. The forest was allowed to
108 recover without intervention (except fire suppression) until 1963. Since then, it has been actively
109 managed using uneven-aged treatment. Major stand entries in Compartment 160 during the monitoring
110 period include: sanitation cutting of mistletoe affected trees, 1982; group selection, 1984; single tree
111 selection, 1995; and thinning, 2005. Regular clearing for fuel reduction and safety near roads impacts
112 only a small portion of the compartment. In 2013, stand basal area averaged 169 ft² ac⁻¹. White fir and
113 Douglas-fir were the dominant species (> 20% relative dominance) but incense-cedar (18%), ponderosa
114 pine (16%), sugar pine (12%), and black oak (9%) were common (Table 1). Between 1983 and 2013,
115 basal area remained relatively stable (3% increase) while tree density declined by 33% (Table 1). During
116 this period, the aboveground live tree biomass averaged 33.5 metric tons ac⁻¹ (\pm 1.8 metric tons ac⁻¹,
117 standard error) and snag biomass averaged 2.8 metric tons ac⁻¹ (\pm 0.6 metric tons ac⁻¹).

118

119 **Snag monitoring.** In 1983, snags \geq 8 in DBH and \geq 4.5 ft tall were inventoried and tagged in
120 Compartment 160 following the protocols established by Raphael and White (1984) at Sagehen Creek

121 Experimental Forest (Sagehen). Specifically, snags were identified to species if possible; otherwise they
122 were identified to genus (e.g., fir, pine, cedar, oak) or by type (i.e., unknown conifer, unknown
123 hardwood). DBH and height were measured along with a list of snag characteristics including wood
124 hardness (scale of 1 to 5), a count of nest holes, limb condition (a categorical class based on the presence
125 of foliage, twigs, and branches), evidence of woodpecker activity (yes or no), broken treetop (yes or no),
126 and percent of stem bark remaining (0-100 %). These characteristics were used by Raphael and White
127 (1984) to quantify the wildlife habitat value of snags.

128 The goal was an inventory of all snags. While an organized search was conducted, it was not a
129 systematic search. As a result, most, but not all, snags were found. Subsequent surveys in 1989, 1994,
130 1995, and 2005 followed the same protocols. In 1994 and 1995, a sample of smaller snags between 4.5 to
131 8 in DBH were tagged and measured in the southeastern subsection of the compartment (approximately
132 18 ac). In 2012, we included all snags ≥ 4 in DBH in the compartment, identified each snag, measured
133 DBH, and summarized snag characteristics using the decay class for standing dead trees from the United
134 States Forest Service Forest Inventory and Analysis field protocol (Section 5.23, United States Forest
135 Service 2022). In 2012, we made every effort to determine the fate of all tagged snags and to capture all
136 newly recruited snags, but we did not establish a systematic process.

137 In 2018, we installed 16 permanent, parallel transects in Compartment 160. They traverse the
138 widest axis of the compartment (east-west) and were spaced 131 ft (40 m) apart along the north-south
139 axis (Fig. 1). All snags in the inventory were located with reference to the east-to-west distance along a
140 specific transect and the north-south offset from the transect centerline. To inventory the entire
141 compartment, searches extended 66 ft (20 m) to the north and south of each transect. Navigation-grade
142 GPS devices record transect distance and offset with sufficient accuracy to reliably relocate snags.
143 We used the transect mapping system to conduct snag inventories in 2018 and 2020, we followed the
144 2012 methods. In 2019, we measured all snags with the combined snaprotocols to create a crosswalk

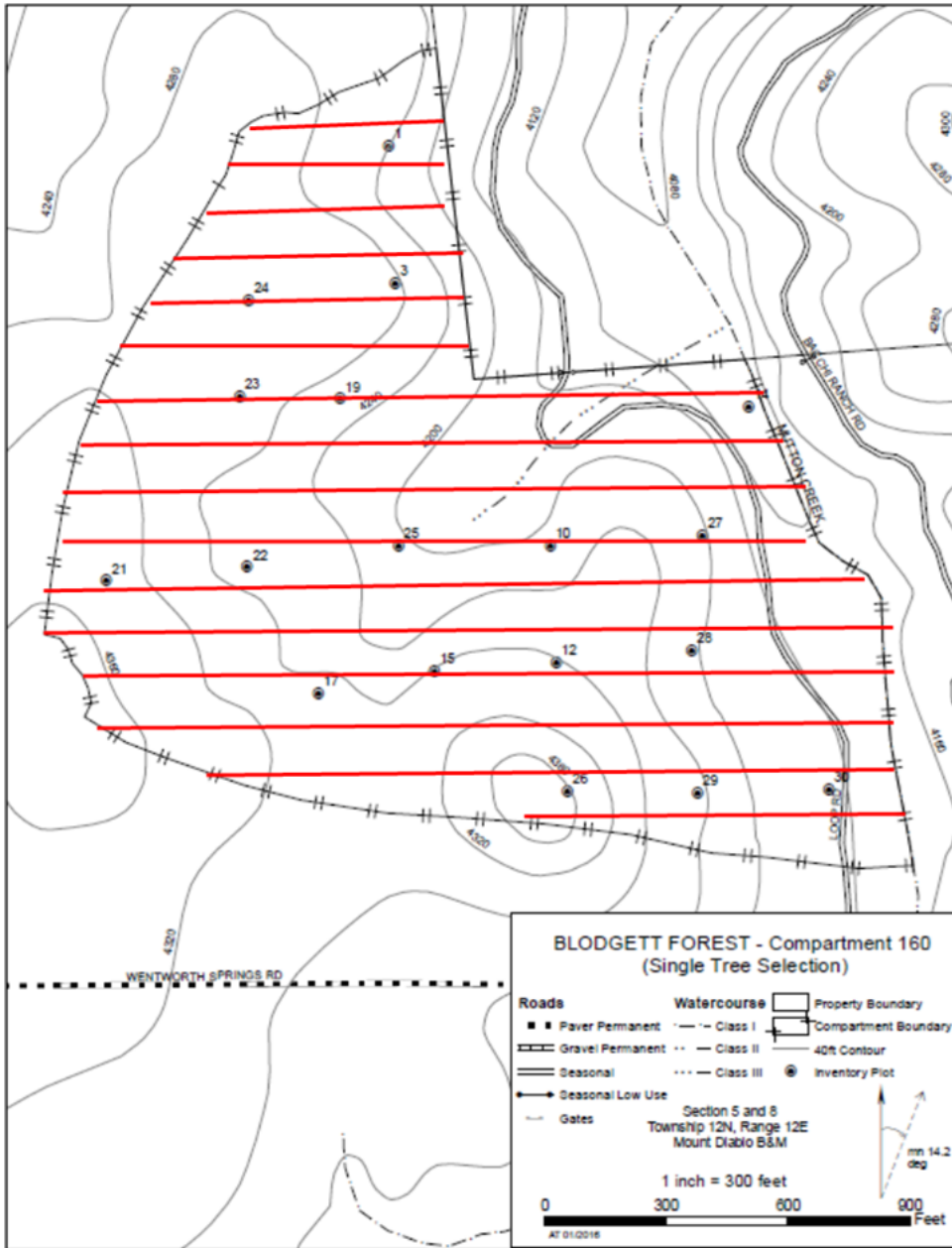
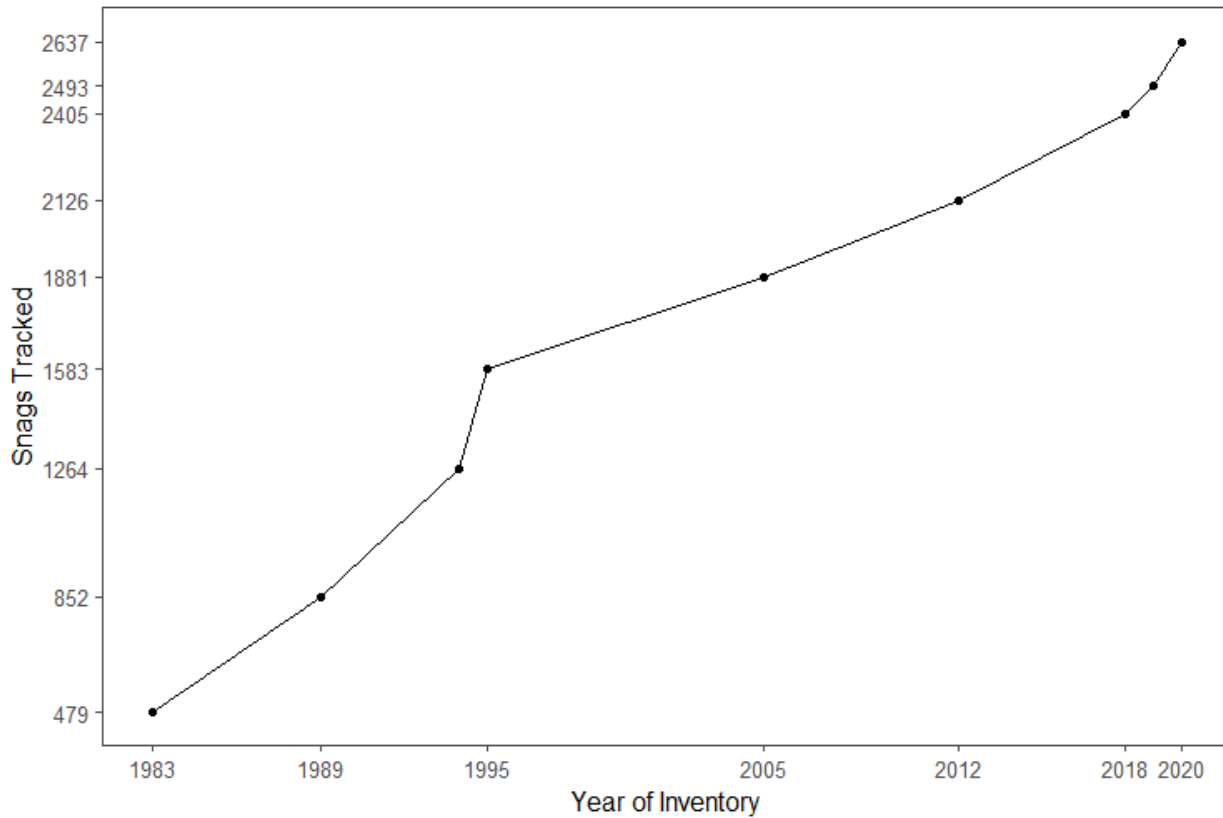


Figure 1. Map of Compartment 160 at Blodgett Forest Research Station. The red lines indicate the location of the snag sampling transects.

147 between the Raphael and White decay characteristics and the FIA decay classes. Beginning in 2019, we
148 also measured snag height and Raphael and White decay characteristics for all newly recruited snags.
149 Currently we are monitoring 747 snags and have tracked a total of 2,378 since the start of this study (Fig.
150 2).

151 **Downed wood decay study.** We established a “common-garden” experiment in Compartment 160 to
152 quantify the effects of species’ wood traits and site-related environmental conditions on wood
153 decomposition rate (sensu Cornelissen et al. 2012). The basic idea is to cut live trees, section the stem
154 into 3-ft lengths, and arrange all the lengths in ordered proximity. The arrangement of the logs resembles
155 the organization of graves at a cemetery (Fig. 3). We included four of the six dominant species: white fir,
156 ponderosa pine, incense-cedar, and black oak; we identified two distinct understory environments: gaps
157 and intact canopy. In total, there were eight common gardens installed; four in each understory
158 environment. We excluded Douglas-fir from the sample to concentrate efforts on less well-studied
159 species. Decay rates for Douglas-fir have more empirical estimates than any other North American tree
160 species (Harmon et al. 2020). Sugar pine was excluded to abide by BFRS’s conservation policy.
161 While our goal was to collect representative samples, typical strategies to ensure success (e.g., random
162 selection) were logistically infeasible for two reasons. One, we could not select trees in the largest size
163 class. Both the commercial value and operational difficulty increase with tree size. Two, an appropriate
164 sample tree for each species needed to be in the vicinity of the common garden. Since all logs had to be
165 moved to the garden manually, distance and terrain were limitations. In response to these limitations, we
166 scouted the entire compartment for possible locations. We found 13 and selected the eight that were most
167 broadly distributed.

168 In October 2018, we felled 32 trees (8 sites; 4 species). Five, 3-ft lengths were cut above the
169 basal swell of the stem. These logs were transported and arranged in the common garden (Fig. 3). After
170 each log was tagged, the length and end diameters were measured along with three estimates of bark
171



172 **Figure 2.** Cumulative number of unique snags tracked in Compartment 160 at Blodgett Forest Research
 173 Station.
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 175



Figure 3. Arrangement of the logs in the long-term wood decay study at Blodgett Forest Research Station. Each row of five logs represents a decay chronosequence for one species.

176 width. To calculate initial wood density, a cross-section of the stem approximately 1 in thick was cut and
177 the “cookie” was transported to the lab. For each cookie, wood density was measured following standard
178 procedures (Cousins et al. 2015). Follow-up log volume and density measurements are planned for 2022,
179 2026, 2032, 2034, and 2038 (i.e., every four years).

180 **Analysis.** We quantified snag fall rates by tracking the fate of recently created snags (i.e.,
181 cohorts) through time. Given the objective of estimating rates, we developed accelerated failure time
182 (AFT) models where the “failure” was defined as snag fall (sensu Hane et al. 2019). We also calculated
183 probabilities of snags to transition from one decay class to the next. These transition probabilities were
184 estimated on an annual basis. Finally, we combined results from our failure time model and transition
185 matrix in a Monte Carlo simulation model to project two management-relevant insights: the expected
186 longevity of snags that provide wildlife habitat and the estimated contribution of snags to downed wood.
187 All analyses were conducted in the R statistical software (R Core Team, 2019).

188 AFTs rely on parametric distributions to directly estimate an annual, baseline failure rate.
189 Covariates, in this case species and size class, effect the baseline rate multiplicatively to either speed up
190 or slow down the failure time. An important complication in this snag cohort study was the uncertainty
191 surrounding the timing of the event for some individuals. Not only did some individuals not fall, i.e. the
192 event occurred beyond the observation period, but also the intervals between measurements varied from
193 one year to nine years. AFTs account for these uncertainties by censoring the observation data: right
194 censored when the event occurs beyond the observation period and interval censored when the event
195 occurs between surveys longer than one year apart (Maringer et al. 2021).

196 We filtered and grouped the data prior to analyses to ensure robust results. First, we only
197 included recently recruited snags in developing the AFT model (N = 722). We defined recently recruited
198 as a snag with an intact stem and limbs, 100% bark cover, and the presence of fine twigs (i.e., FIA decay
199 class = 1). Second, we combined ponderosa pine and sugar pine into one generic pine species to increase

200 sample size. The two pine species had similar fall rates in a prior analysis (Battles et al. 2015). We also
201 added the few Douglas-fir snags (N=23) into the generic fir species group. Third, we considered DBH
202 size classes, as defined by wildlife usage (Raphael and White 1984), as the most informative size
203 covariate. In particular, Raphael and White (1984) documented the low usage of small snags (DBH < 9
204 in) by cavity-nesting birds.

205 We used the `icenReg` package for R (Anderson-Bergman 2017) to develop fall rate regressions.
206 First, we evaluated the fit of a suite of parametric distributions (e.g., exponential, gamma, and Weibull)
207 to the semi-parametric proportional hazard model of snag fall. We then used an information theoretic
208 approach to select the best AFT model (Burnham and Anderson 2002) from five possible combinations
209 of covariates: null (no covariates, intercept only), species group only, DBH class only, additive effects of
210 species group and DBH class, and interactive effects of species group and DBH class. The criteria for
211 model selection were based on the Akaike Information Criterion (AIC). For this set of potential models,
212 we calculated the difference in AIC between every model and the model with the lowest AIC (Δ AIC)
213 and the weight of evidence for each model.

214 We quantified the progression of snag decay by calculating persistence rates from observed
215 changes in FIA decay class. These rates were based on the persistence of snags in their initial decay class
216 across each survey interval. This approach yielded a total of 2,202 observations that spanned decay
217 classes. Specifically, there were 891 snags with an initial decay class of 1; 543 snags with a decay class
218 of 2, 211 snags with a decay class of 3, and 557 with a decay class of 4. Given the subjective nature of
219 assigning decay class, there were instances of “regression” where over time the snag moved from a class
220 indicating more decay to one indicating less decay. Regression of more than one class was rare. One-
221 class regressions were most common between class 3 and 4. Since both decay class provide similar
222 wildlife habitat elements (e.g., nest cavities, Raphael and White 1984), we combined them into one class
223 (3/4). When calculating persistence, we considered instances of regression as evidence of persistence in

224 the initial decay class. In other words, if a snag was assigned a decay class = 2 in the initial survey and a
225 decay class = 1 in the follow-up survey, we counted this snag as persisting in the initial decay class.
226 Annual rates of persistence were estimated for each species group and decay class group. Since the
227 lengths of the survey intervals varied, we divided interval persistence by the number of years between
228 surveys. Overall persistence rates were estimated using the weighted average (weights = total number of
229 snags) of the interval results.

230 To gain management relevant insights from snag fall rates and decay class persistence, we
231 developed a Monte Carlo simulation that predicts the fate of 722 recently recruited snags (i.e., FIA decay
232 class = 1) in our dataset. Using our best AFT snag fall model, we imputed (icensReg package, Anderson-
233 Bergman 2017) 1,000 realizations of fall times for each of the 722 trees. Each simulation starts with a
234 random selection of the fall time for every snag. We then used the persistence rates to predict transitions
235 through the decay classes for the years that the snag remains standing. The simulated, yearly decay
236 progression begins with a randomly generated persistence probability. If the observed persistence rate is
237 greater than or equal to the random probability, the snag remains in its decay class. If the observed
238 persistence rate is less than the random probability, the snag progresses to the next decay class. The
239 cycle proceeds through the decay classes using the appropriate observed persistence rate for each class
240 until the simulation reaches the fall time. Results from these 1,000 simulations were summarized with
241 medians and interquartile (i.e., 25th -75th percentile) ranges.

242 For each step in the Monte Carlo simulation, we calculated the stem biomass of each snag and
243 expressed it relative to the mass of a live tree of the same species, DBH, and height (sensu Harmon et al.
244 2011). We used regional biomass equations (Forest Inventory and Analysis 2010) to estimate live stem
245 biomass. We then discounted stem biomass based on decay class using results from Cousins et al. (2015)
246 for all the conifers and results from Harmon et al. (2011) for the hardwoods. We converted biomass to

247 carbon using the mean carbon density for snags reported in Cousins et al. (2015), namely 0.523 grams of
248 carbon per gram of biomass.

249

250 RESULTS

251 The best ACF model of snag fall at BFRS used the Weibull distribution and included the additive
252 effects of species group and DBH class. The Weibull distribution best approximated the non-parametric
253 Turnbull estimator (Fig. S1). The Turnbull estimator is a generalized version of the Kaplan-Meier
254 estimator that can handle censored data (Anderson-Bergman 2017). Among the Weibull-based ACFs, the
255 additive model with both covariates had the most support ($\Delta AIC = 3.8$; weight of evidence = 0.85, Table
256 S1). For this cohort of BFRS snags, the median time to fall was 14 years (interquartile range = 8-21
257 years). As a test of the fit of the best ACF model, we also calculated the median fall time using the
258 Turnbull estimator (sensu Maringer et al. 2021). There was good agreement between the parametric and
259 non-parametric results – the median fall time using the Turnbull estimator was also 14 years.

260 The species effect was driven by slower fall rates for incense-cedar compared to other species
261 groups; the size effect was driven by slower fall rates for the largest size class (Fig. 4, Table S2).
262 Specifically, the median fall time for incense-cedar across size classes was 21.1 years. For the three other
263 species groups, comparable median fall times ranged from 13.8 years for fir to 15.3 years for pine.
264 Although median fall times steadily increase by size class, the biggest increase was from the DBH = 15-
265 21 in class where the median fall time was 16.4 years to the DBH > 21 in class where the median fall
266 time was 20.2 years (an increase of 3.8 years). As indicated by the strength of the additive model,
267 species and size effects were consistent across all species by size class pairings (Fig. 5). However, there
268 was considerable uncertainty in the projected median fall rates. The interquartile ranges spanned 90% of
269 the median fall times.

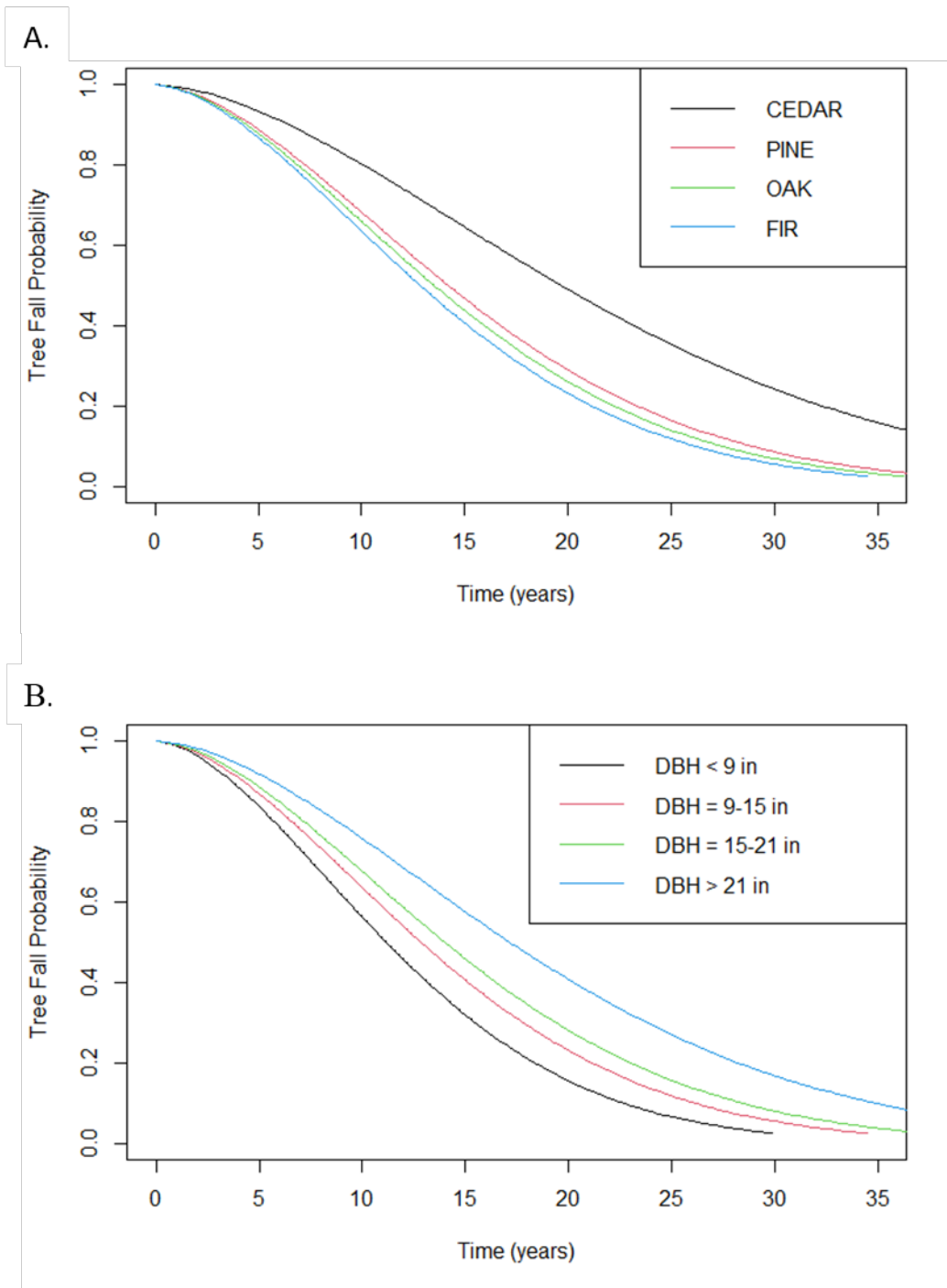
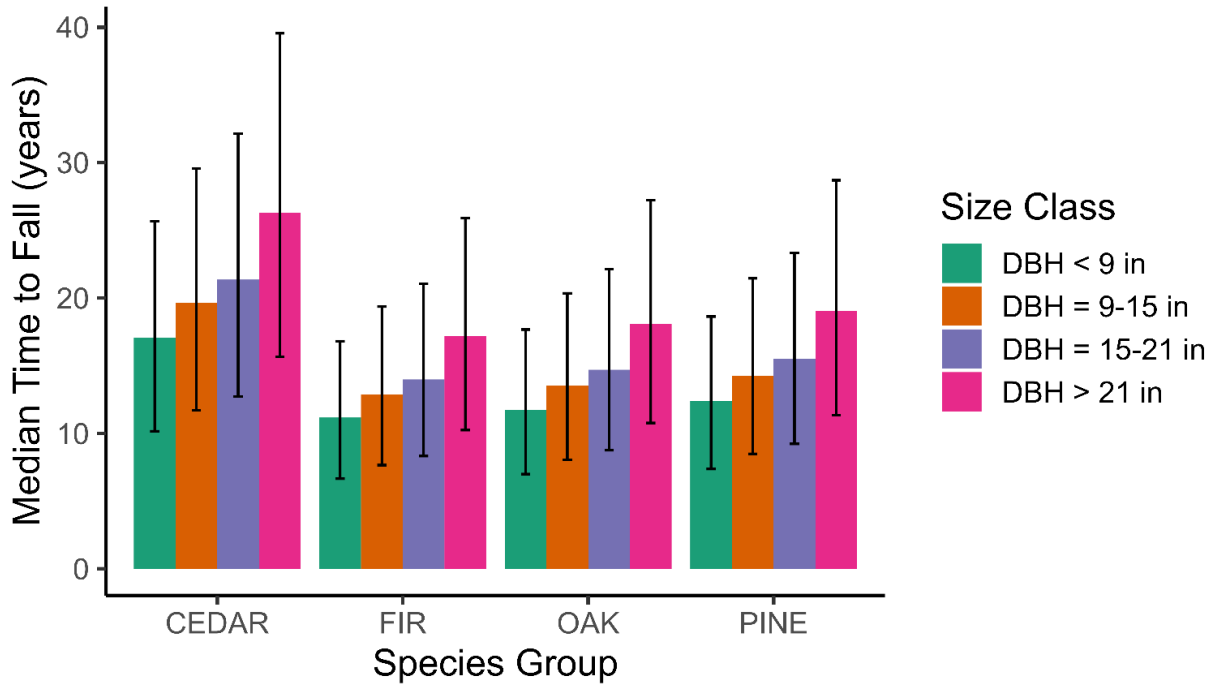


Figure 4. Accelerated failure time models for tree fall probability at Blodgett Forest Research Station. A. Differences in tree fall probability by species group. Results for snags with DBH = 9-15 in. B. Differences in tree fall probability by DBH size class. Results for Fir species group.



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274 **Figure 5.** Predictions of median fall time for snags at Blodgett Forest Research Station by species group
 275 and size class. The error bars represent the interquartile range.

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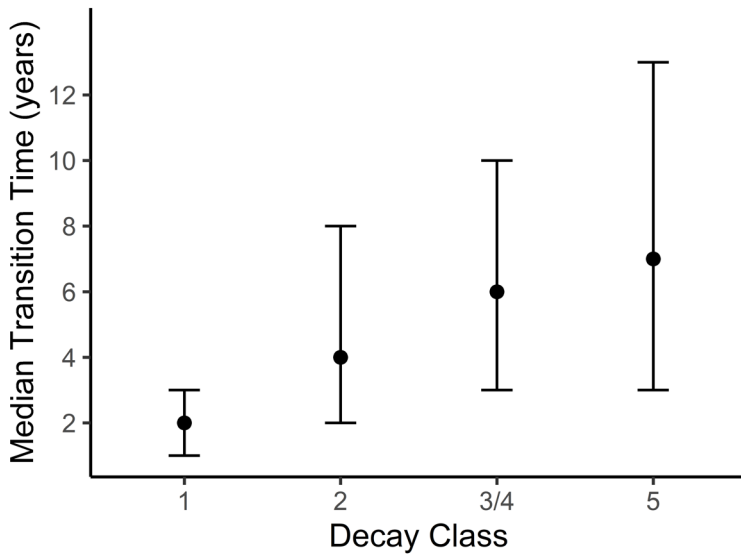


Figure 6. Predicted transition time in each decay class for snags at Blodgett Forest Research Station. Error bars represent the interquartile range. Results from Monte Carlo simulation of fall probabilities and decay rate

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278 For snags that remained standing, incense-cedar also had the slowest decay transition rate at
279 every stage of the process (Table 2). Of the remaining species group, pine snags transitioned the slowest.
280 Probability of transitioning was the highest from decay class 1 to decay class 2 (T1) and lowest from
281 decay class 3/4 to decay class 5 (T3/4). For example, the average T1 transition probability was 0.33 yr^{-1}
282 compared to 0.06 yr^{-1} for T3/4 transitions – a 5.5x difference.

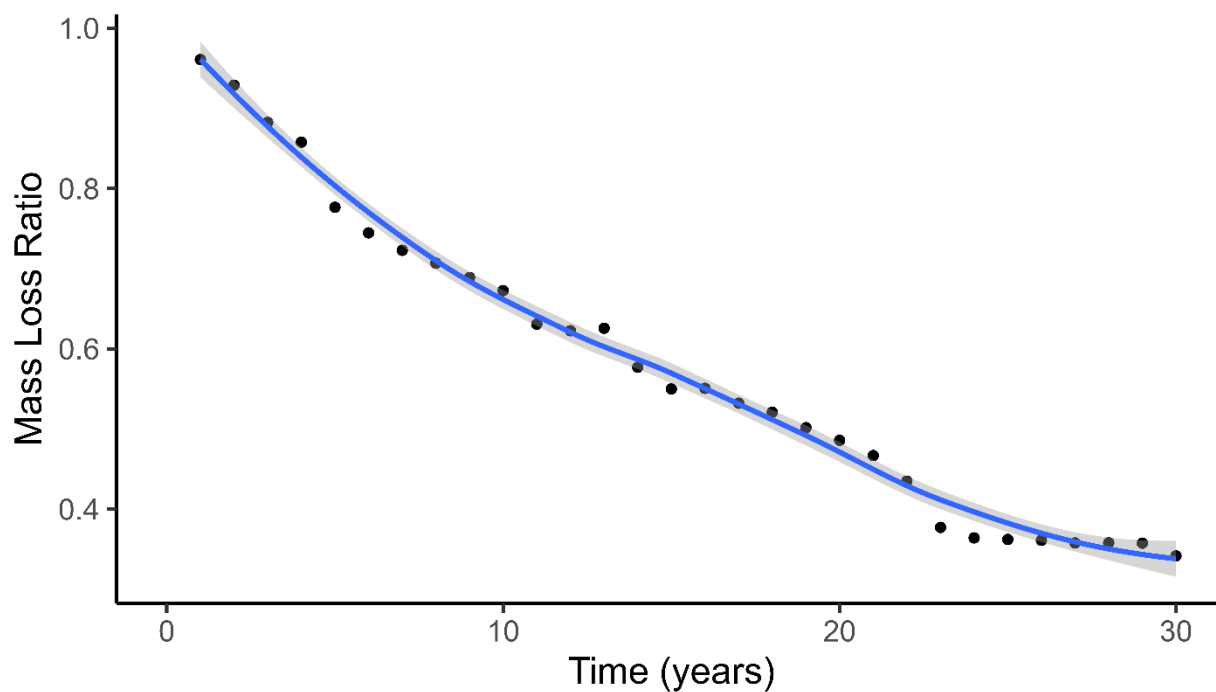
283 Our Monte Carlo simulation of the fate of the 722 newly created snags at BFRS produced four
284 important insights. 1) While the median fall time for snags was 14 years, the rate varied with time since
285 death. There was an initial lag in snag fall. On average, only 15% of the cohort fell in the first five years
286 (108 ± 5 snags, mean \pm standard error). Rates accelerated between year 5 and 15 (42% of the cohort fell)
287 and then slowed precipitously to 7% between year 25 and 30. After 30 years, the fall rate flattened to less
288 than 2% of the cohort falling every five years. 2) The mean annual probability of falling ($9\% \text{ yr}^{-1}$) for
289 decay class 1 snags was two-to-four times lower than the other decay classes. Snags in decay class 3/4
290 had the highest fall probability with a mean equal to $39\% \text{ yr}^{-1}$. 3) For snags in decay classes that provide
291 essential wildlife habitat (i.e., decay classes 2 and 3/4), the median persistence time was 10 years (Fig.
292 6). However, the interquartile range for persistence in a decay class was large and the distributions of
293 persistence times had long tails. For example, the 75th percentile of persistence for decay class 2 snags
294 was eight years and for decay class 3/4 it was 11 years. 4) As snags decay, they collectively lost mass at
295 an exponential rate (Fig. 7). After 14 years when half of the snags had fallen, the remaining snags
296 retained only 58% of their original mass. Thus after 14 years, 70% of the carbon in new snags has either
297 been transferred to the dead and downed wood pool (snags that fell) or respired to the atmosphere as CO_2
298 (wood that decayed).

299 The 32 felled trees used to establish the log decay study ranged in DBH from 8 to 18 in with a
300 median of 11 in. All but one was large enough to provide wildlife habitat. Series of logs approximately

Table 2. The annual transition probability (%) for snag decay classes from Blodgett Forest Research Station. Note: Decay class 5 does not have a transition probability since it represents the most decayed state.

	T1	T2	T3/4
Cedar	0.22	0.12	0.03
Fir	0.45	0.15	0.07
Oak	0.35	0.17	0.07
Pine	0.30	0.11	0.07

301



302

303 **Figure 7.** Trends in the loss of stem mass snags at Blodgett Forest Research Station. Results from Monte
 304 Carlo simulations of fall probabilities and decay rate. The black circles represent the mean mass loss
 305 ratio (snag stem mass:live stem mass) for remaining snags. The blue line is the locally fit smoothing
 306 spline; the gray shading represents the 95% confidence interval.

307

308

309 three feet long were arranged in the two understory environments, specifically in gaps and in the forest
310 matrix. The completed set-up was a replicated ($n = 4$), full-factorial experiment (species by
311 environment). For all 160 logs, we measured the length of the log, the diameters at both ends, and
312 average bark width from three bark samples. We also quantified the live wood density of each felled tree
313 (Table 3). Black oak had the densest wood (oven-dry density $> 39 \text{ lbs ft}^{-3}$) while white fir and incense-
314 cedar were the least dense (oven-dried density $< 27 \text{ lbs ft}^{-3}$).

315

316 MANAGEMENT RELEVANCE

317 The goal of this project was to quantify the life cycle of standing dead trees (i.e., snags) to inform
318 forest management and policy development. The innovative snag research at Sagehen Creek
319 Experimental Forest (Raphael and White 1978, Raphael and Morrison 1987, Morrison and Raphael
320 1993) provided the empirical foundation for our study. Raphael and White (1978) quantified the utility
321 of snags as wildlife resources; Raphael and Morrison (1987) documented the interdependence of snag
322 decay and snag fall; and Morrison and Raphael (1993) implemented a predictive model of snag
323 dynamics. BFRS managers followed the key recommendations from the Sagehen research. They
324 designed a study to understand how to conserve snag-dependent wildlife habitat in managed forests
325 (Raphael and White 1979) and to test the applicability of the Sagehen snag dynamic models to other
326 geographic areas (Morrison and Raphael 1993). While the results from BFRS support key aspects of the
327 Sagehen results, namely that tree size and tree species are important determinants of snag fall, Raphael
328 and Morrison (1987) reported higher fall rates for newly created snags. They estimated that 40% of the
329 pine snags in the unburned forest fell after five years and that half of all snags fell by 10 years (Raphael
330 and Morrison 1987). By comparison, we projected that only 7% of the pine snags fell after 5 years and
331 that it took 14 years for half of all snags to fall. Since the decay class transition rates for BRFS and

Table 3. Wood density for logs in the long-term decay study at Blodgett Forest Research Station. Mean and standard deviation (sd) reported for each understory environment (n = 4).

Species	Understory Environment			
	Matrix (lbs ft ⁻³)		Gap (lbs ft ⁻³)	
	mean	sd	mean	sd
White fir	26.4	2.0	24.1	2.5
Incense-cedar	26.3	1.9	26.8	3.6
Ponderosa pine	35.4	7.7	30.0	2.2
Black oak	39.2	1.9	39.6	1.3

333 Sagehen are comparable (Table 2, Morrison and Raphael 1993), the early lag in tree fall at BFRS drives
334 differences in the longevity of snags. Our estimate of the median persistence of snags (Fig. 6) in decay
335 classes that provide wildlife benefits (i.e., decay classes 2, 3, and 4) of 10 years was about five years
336 longer than projections based on Morrison and Raphael (1993).

337 The BFRS study also quantified the variation around expected results. For example, the
338 interquartile range for the longevity of wildlife snags was 5-19 years (Fig. 6). Part of this wide range was
339 the result of year-to-year differences in snag fall. For example, of the 13 new snags in the 2018
340 inventory, 54% were down in 2019. In contrast, of the 82 new snags in the 2019 inventory, only 5%
341 were down in 2020. Episodic events (e.g., heavy snowfall, high winds) can lead to pulses of snag fall. A
342 multi-decadal study is more likely to sample these episodic events and thus incorporate their impact into
343 snag dynamics. However, part of the reported variation was due to the study design. The multi-year
344 intervals between surveys introduced uncertainty into timing of snag falls. We were able to formally
345 account for this uncertainty with interval censoring, but the “cost” was increased variation in the
346 predicted snag fall rate (Fig. 5, Fig. S1).

347 The variation noted above in snag fall between the 2018 and 2019 surveys also illustrates the
348 variation in snag recruitment due to differences in annual tree mortality rates. Based on plot inventories
349 in Compartment 160 at BFRS, annual mortality has ranged from a high of 2.2 % yr⁻¹ between 1994-1995
350 and a low of 0.13 % yr⁻¹ between 2003-2005. Raphael and Morrison (1987) incorporated the production
351 and loss of snags in their population dynamics model. They emphasized the need to consider the entire
352 life cycle of snags.

353 The uneven-aged management of Compartment 160 at BFRS has maintained an ample supply of
354 snags with snag fall largely replaced by snag recruitment. The density of snags providing wildlife habitat
355 (i.e., snags with DBH > 9 in) was 15 stems ac⁻¹ in 1983 and 13 stems ac⁻¹ in 2013. At Sagehen during a

356 period with no active management, snag density on unburned plots increased from 7 stems ac^{-1} in 1978
357 to 9 stems ac^{-1} in 1988 (Morrison and Raphael 1993). However, more intense management strategies
358 may require more explicit planning. For example, the recently completed (2018) fuel reduction
359 treatments at Sagehen decreased pre-treatment snag density by almost 50% with reductions in all size
360 classes (Manley et al. 2022). The snag losses at Sagehen were mainly the result of operational necessity.
361 They were removed to safely complete the thinning and mastication treatments (Manley et al. 2022). To
362 maintain large snags (DBH > 15 in) at a density of 1-3 snags ac^{-1} in areas designated to support wildlife
363 habitat, the treatment plan included the partial girdling of live trees (United States Forest Service 2013).

364 To maintain snag populations under even-aged management regimes similar planning for
365 “decadent features” may be necessary. For productive, California mixed conifer forests (i.e., sites
366 comparable to BFRS), results from this study provide some useful guidelines:

367 Snags in the early stage of decay typically provide critical wildlife habitat elements for
368 approximately 10 years (Fig. 6).

369 The persistence of snags increases with DBH and for incense-cedar (Fig. 5). These two
370 factors can double the median longevity of snags.

371 It takes from one to three years for a newly created snag to degrade and decay to the point
372 where it provides wildlife habitat value (Fig. 6).

373 Cavity-nesting birds only use snags with a DBH of least 9 in (Raphael and White 1984).

374 Old-growth dependent mammals like the pine marten and pacific fisher only use snags in the
375 largest size class (Spencer 1987, Purcell et al. 2009).

376 These guidelines must be applied with appropriate caution. Raphael and Morrison (1987) noted that the
377 fall rate can vary with the cause of death. In particular, fire killed snags tend to fall faster than trees that
378 died from other causes (Grayson et al. 2019). Trees killed by beetle attack also may fall faster than trees
379 that died from competitive stress (Keen 1955). Differences in soil and canopy structure can also
380 influence the survival times of snags (Grayson et al. 2019).

381 The Sagehen snag management plan (United States Forest Service 2013) was informed by
382 predictions from the snag submodel in the Fire and Fuels Extension (FFE) to the Forest Vegetation
383 Simulator (FVS, Rebain 2022). The more detailed and site-specific model for snag dynamics (Morrison
384 and Raphael 1993) was not used. FFE has proven to be a vital addition to the widely used FVS. However
385 key elements of FFE were built on the best-available, but sparse, information. Specifically for the
386 western Sierra Nevada forests (WS variant), the FFE snag fall model is based on unpublished data with
387 species-specific parameters developed by the variant user group (Rebain 2022). For snags created by
388 fire, Grayson et al. (2019) noted that FFE predicted higher 10-year snag survival than their field
389 observations. Given the ubiquity of planning tools like FVS, it is essential to incorporate the empirical
390 insights from these snag survival studies into commonly used modeling framework. It is unlikely that the
391 typical forest manager would find the model predictions in Table S2 much help given constraints on time
392 and training.

393 Cousins et al. (2015) documented the decay trajectory of snags in the mixed conifer forest. In
394 terms of carbon dynamics, more than 40% of the carbon was lost as a snag progresses through decay
395 classes. When this decay is combined with transition to the dead and downed wood pool, the ephemeral
396 nature of carbon stored in snags is striking (Fig. 7). Given that much of the snag carbon becomes downed
397 and dead wood carbon, understanding the rate of log decay is essential. The log decay experiment
398 established at BFRS will help fill this gap in our knowledge.

399 The BRFS and Sagehen studies illustrate three unique challenges in monitoring snag dynamics.

400 1) In absence of a major disturbance, snags are a minor component of standing stems in a forest. Thus, a
401 large area must be searched to include sufficient samples. 2) Snag fall rates are not necessarily constant.
402 Our results included an early lag and a late tail in fall rates (Fig. 4). To capture initial dynamics, annual
403 sampling is needed. To capture long-term dynamics, snags must be tracked for decades. 3) Yet, snags
404 can be hard to track. Over time, a snag loses its distinguishing features like crown structure, bark,
405 diameter, and height. Often the tag is the only means to a positive identification but as wood decays,
406 keeping a tag in a snag becomes more difficult. And finding a tag on a fallen, disintegrated snag can
407 require an extensive search.

408 At Sagehen (Raphael and Morrison 1987, Morrison and Raphael 1993) applied proven forest
409 inventory techniques. They tracked snags with fixed area plots, tagged trees, and periodic censuses
410 (every five years). Their initial sample included 774 snags in five, unburned 21-acre plots (Morrison and
411 Raphael 1993). For live trees, five years is a reasonable inventory interval, but for snags, it misses the
412 initial dynamics. Moreover, fixed plot inventories are a considerable investment that can prove hard to
413 maintain over the long-term. The Sagehen study ended after 10 years. The early snag surveys at BRFS
414 captured enough snags with modest effort but they were not systematic enough to reliably track the fate
415 of all snags. Too many tagged snags were not found on subsequent surveys. The system we installed
416 provides a spatial framework for the survey. The transects along with the approximate locations based on
417 hand-help GPS devices narrowed the search area sufficiently so the recovery rate of tagged snags
418 exceeded 99%. This systematic search approximately doubled the labor costs. Based on the 2019 and
419 2020 remeasurements, it takes a three-person crew two weeks to complete the field work (240 person-
420 hours).

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Table S1. Model covariates, AIC scores, and weight of evidence for the top five accelerated failure time models. Additive models denoted with a “+”; models with interactions denoted with a “*”; k is the number of parameters in the model.

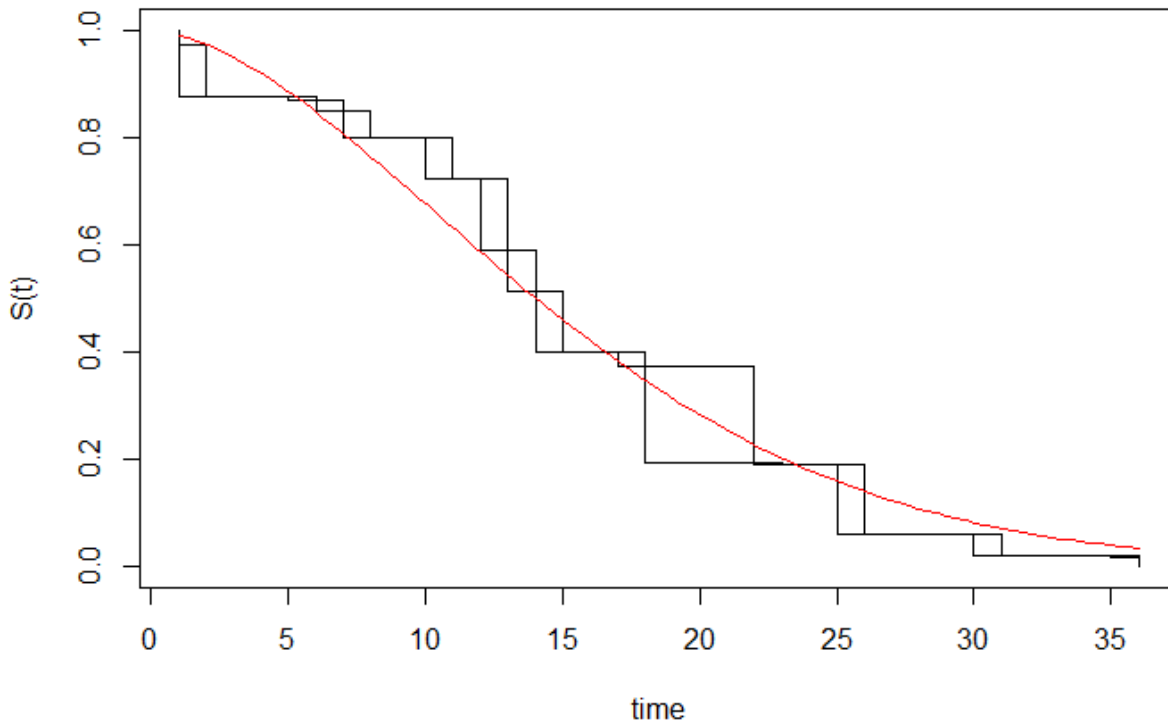
Covariates	Distribution	k	AIC	ΔAIC	weight of evidence
SppGrp+DBHbin	Weibull	9	1124.6	0.0	0.85
SppGrp	Weibull	6	1128.3	3.8	0.13
SppGrp*DBHbin	Weibull	18	1132.7	8.2	0.01
null	Weibull	3	1136.4	11.8	0.00
DBHbin	Weibull	6	1137.8	13.2	0.00

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Table S2. Predictions of snag fall times by species group and size class. IQ is the interquartile (25th – 75th percentile) range.

Species Group	Size Class (DBH range)	Fall Time (years)	
		Median	IQ Range
CEDAR	DBH < 9 in	17	10 - 16
CEDAR	DBH = 9-15 in	20	12 - 30
CEDAR	DBH = 15-21 in	21	13 - 32
CEDAR	DBH > 21 in	26	16 - 40
PINE	DBH < 9 in	12	7 - 19
PINE	DBH = 9-15 in	14	8 - 21
PINE	DBH = 15-21 in	15	9 - 23
PINE	DBH > 21 in	19	11 - 29
OAK	DBH < 9 in	12	7 - 18
OAK	DBH = 9-15 in	14	8 - 20
OAK	DBH = 15-21 in	15	9 - 22
OAK	DBH > 21 in	18	11 - 27
FIR	DBH < 9 in	11	7 - 17
FIR	DBH = 9-15 in	13	8 - 19
FIR	DBH = 15-21 in	14	8 - 21
FIR	DBH > 21 in	17	10 - 26

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 519 **Figure S1.** Comparison between semi-parametric Turnbull estimate (black lines) of snag survival rate
 520 and the best accelerated failure time model (Weibull distribution, red line) of snag survival. $S(t)$ is the
 521 snag survival probability and time is in years.

522