

1 **Title:** Drought Induced Snag Dynamics and Fuel Succession in the Sierra Nevada

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3 **Highlights:**

4 • Sierra Nevada forests currently contain large pools of combustible biomass in standing  
5 dead trees following extensive drought mortality

6 • As a result of tree mortality, fine woody debris and litter and duff fuel loads increased  
7 significantly from 2017 to 2021

8 • Snag longevity increased with tree size and varied by tree species, with incense-cedar  
9 snags standing the longest and yellow pine snags (ponderosa pine and Jeffrey pine)  
10 standing the shortest

11 • Fuel loads are projected to increase sharply in the coming decades as remaining standing  
12 dead trees fall to the surface

13 **Abstract:**

14 California recently experienced one of the most severe periods of drought in its recorded history  
15 from 2012-2016, which has been followed by widespread tree mortality throughout the state with  
16 estimates suggesting over 147 million trees have died. One of the most concerning effects and  
17 uncertainties surrounding the episodic mortality is the sudden creation of millions of standing  
18 dead trees (snags), sharply increasing snag biomass in Sierra Nevada forests. Despite the  
19 importance of snags in influencing wildlife habitat, nutrient cycling, and wildfire effects, the  
20 dynamics of snag persistence and decay rates are still not particularly well understood and  
21 characterized. Here we attempt to better characterize the likely progression of fuels and snag  
22 decay following the recent extensive mortality event in the Sierra Nevada mixed conifer  
23 ecosystem using long term annually collected snag demography data and recent surface fuel

24 measurements from three locations. In the short term (2017-2021) fine woody debris and litter  
25 and duff significantly increased across all three study sites, while coarse woody debris increased  
26 only at one site, and total fuel loads increased at two of the three sites. Across all species snag  
27 longevity increased with size with the relationship varying by species. Size played a small role in  
28 influencing yellow pine (ponderosa pine and sugar pine) fall rates but a very large role in red fir  
29 and sugar pine longevity. Size played an important, but less pronounced role for white fir,  
30 incense-cedar, and black oak. Our projections show that the rate of biomass input from future  
31 snag fall varies across our three sites and by 2040 we project median cumulative inputs of 136.1  
32 Mg ha<sup>-1</sup>, 92.3 Mg ha<sup>-1</sup>, and 49.4 Mg ha<sup>-1</sup> at each site.  
33 Managers are not without options for how to deal with the extreme number of snags and their  
34 large ground combustible biomass. Careful timing and application of prescribed burning and  
35 salvage operations may effectively reduce fuel loads and change fuel load trajectories to limit  
36 fire risk into the future, but both methods have operational, economic, and legal limitations. This  
37 study provides managers with some essential information regarding snag longevity and fuel load  
38 accumulation following the extensive drought in the Sierra Nevada that can aid in future decision  
39 making.

40

41 **Keywords:** California, snag fall rates, fuel loads, bark beetle mortality, forest management

42

### 43 **1. Introduction**

44 California recently experienced one of the most severe periods of drought in its recorded  
45 history (Griffin & Anchukaitis, 2014; Robeson, 2015), which has been followed by widespread  
46 tree mortality throughout the state with estimates suggesting over 147 million trees have died

47 (USDA 2019). Due to over a century of fire suppression and exclusion without concurrent  
48 density management, many Sierra Nevada forests existed at high densities prior to the drought  
49 and were likely under much higher competitive stress relative to their pre-European historical  
50 conditions (Collins et al., 2011; North et al., 2022). High levels of competition coupled with  
51 stress due to five years (2012-2016) of extreme drought limited the ability of trees to resist  
52 attacks from insects and pathogens. Ultimately many of these drought-weakened trees died (Kolb  
53 et al., 2016; Stephenson et al., 2019). Numerous field-based studies have characterized the  
54 immediate impacts of the drought, noting extensive mortality of overstory trees throughout the  
55 Sierra Nevada, particularly in the southern region of the range (Axelson et al., 2019; Fettig et al.,  
56 2019; Pile et al., 2019). However given that the frequency of droughts of similar magnitudes are  
57 projected to increase in the future (Diffenbaugh et al., 2015), it is important to better understand  
58 the longer-term impacts of massive tree mortality on forest dynamics. .

59         An immediate concern is that millions of standing dead trees (snags) greatly increase the  
60 amount of dead biomass in Sierrian forests. Snags play important roles in many ecosystem  
61 processes. They provide important habitat for many cavity-nesting birds and wildlife while  
62 standing and can house many invertebrate and fungi species while standing or after falling to the  
63 ground (Harmon et al., 1986; Raphael & White, 1984; Saab et al., 2014). The decomposition of  
64 snags and the coarse woody debris created when snags fall (CWD) also plays a critical role in the  
65 cycling of nutrients and carbon within ecosystems (Cousins et al., 2015; Harmon et al., 1986).  
66 Snags also influence wildfire risk and fire behavior. Burning snags can produce embers that  
67 facilitate spot fires (Barrows 1951) while decayed snags and coarse woody debris on the ground  
68 can provide a receptive surface for spot fires (Knapp, 2015; Stephens, 2004). Furthermore, the  
69 combustion of coarse woody material can result in long burnout times that release high amounts

70 of energy and heat to the soil, increasing severity and effects of fire (Brown et al., 2003;  
71 Monsanto & Agee, 2008). More recently, the amount of coarse woody material and density of  
72 snags has been associated with high fire severity and mass fires (Jaffe et al., 2021; Lydersen et  
73 al., 2019; Stephens et al., 2018).

74         Despite the importance of snags in influencing wildlife habitat, nutrient cycling, and  
75 wildfire effects, the dynamics of snag persistence are not well-characterized, particularly in  
76 response to catastrophic mortality events other than wildfire. Studies have identified common  
77 traits that influence fall rates, such as size, species, height, and site characteristics (Audley et al.,  
78 2021; Parish et al., 2010; Raphael & Morrison, 1987; Rhoades et al., 2020), often reporting that  
79 snag longevity varies based on species and that larger trees tend to fall at slower rates. However,  
80 many studies focus specifically on trees that were killed by wildfires (Grayson et al., 2022;  
81 Ritchie et al., 2013) or report on fall rates in different ecosystems than the Sierra Nevada  
82 (Audley et al., 2021; Parish et al., 2010; Rhoades et al., 2020). Despite research on non-fire  
83 killed snags in the Sierra Nevada (Raphael & Morrison, 1987), fall rates are rarely studied in a  
84 manner that allows managers and scientists to accurately predict the longevity of a given snag.  
85 Furthermore, it is not well understood exactly how much dead overstory trees contribute and  
86 relate to surface fuel loads over time (Lydersen et al., 2015). With over 147 million dead trees  
87 across Sierra Nevada forests, it is essential to understand the potential timing and magnitude of  
88 fuel inputs from snags in the coming years as millions of snag decay and fall to the surface.

89         In the wake of the recent drought-mortality event in the Sierra Nevada, the amount of  
90 combustible woody material on the surface floor is expected to dramatically increase as snags  
91 decay and fall to the ground, ultimately increasing the potential for mass fires (Stephens et al.,  
92 2018). A general model from Hicke et al. (2012) describes multiple phases of fuel and fire risk

93 development following waves of bark beetle mortality, with an initial “red phase” as recently  
94 dead canopy fuels dry out and lose moisture, followed by a “gray phase” as fine foliage and  
95 branches fall to the ground, shifting fine fuels from the canopy to the surface fuel layer, and then  
96 an “old phase” as entire boles and snags fall to the ground, notably increasing surface fuel loads  
97 due to an increase of coarse woody debris from fallen snags. It is hypothesized that the Sierra  
98 Nevada will follow a similar trend in wake of the recent mortality.

99         A number of studies have observed pulses in fine woody material and litter shortly after  
100 bark beetle driven mortality epidemics (Klutsch et al., 2009; Page & Jenkins, 2007; Schoennagel  
101 et al., 2012). Similarly, studies have either directly observed large increases in coarse woody  
102 material as stands enter the “old phase” (Page & Jenkins, 2007; Schoennagel et al., 2012) or  
103 make similar projections of coarse woody debris loads (Collins et al., 2012; Klutsch et al., 2009).  
104 While it is generally understood that there should be a contribution to surface fuel loads as snags  
105 eventually decay and fall over and deposit woody material to the ground, the exact timing of fall  
106 rates and magnitude of fuel inputs are difficult to quantify and vary widely across different  
107 ecosystems.

108         Here we attempt to better characterize the likely progression of fuels and snag decay  
109 following the recent mortality event in the Sierra Nevada mixed conifer ecosystem. We leverage  
110 39 years of annual inventory data and recent surface fuel measurements to quantify short- and  
111 long-term trends in surface fuel changes. We also use accelerated failure time models and recent  
112 field data to make projections for when snags are likely to fall in the coming decades in  
113 Yosemite and Sequoia-Kings Canyon National Parks. The objectives of this study were to 1)  
114 determine the short-term changes in fuel loads over the past 5 years in the central and southern  
115 Sierra Nevada, 2) develop models describing snag fall rates for common Sierra Nevada species,

116 and 3) to apply those models to make projections quantifying the magnitude and timing of  
117 biomass inputs from snag-fall to surface fuels into the future.

118

## 119 **2. Methods**

### 120 *2.1 Study Sites*

#### 121 *2.1.1 Site Description for Fuel Load Plots*

122 We sampled fuels data from a network of plots installed in 2017 across the Sierra Nevada  
123 to research effects of the 2012-2016 drought mortality (Axelson et al. 2019). For this study, we  
124 focus on the three study areas in the Southern Sierra Nevada where drought-induced tree  
125 mortality was most extensive (Fettig et al., 2019; Preisler et al., 2017).

126 The Sequoia-Kings Canyon National Parks (SEKI) site is located in the Crystal Cave area  
127 of Sequoia National Park in Tulare County, CA. The 1700-ha study area is managed by the  
128 National Park Service and has experienced neither fire nor timber harvests in the last century.  
129 Mean elevation is 1687 m. Average annual temperature is 7 °C (January: 0 °C; July: 17 °C) and  
130 average annual precipitation is 1143 mm (PRISM Climate Group, 2020). White fir (*Abies*  
131 *concolor*, Gord & Glend) and incense-cedar (*Calocedrus decurrens*, Torr) dominate the site.  
132 Although situated near old giant sequoia (*Sequoiadendron giganteum*, Buchholz) groves, no  
133 giant sequoia were documented within the study area.

134 The Yosemite Mixed-Conifer (YOMI) site is situated near the Hodgdon Meadow  
135 Campground at the Big Oak Flat Entrance station in Tuolumne County, CA. The 260-ha study  
136 site is managed by the National Park Service. Mean elevation is 1422 m. Average annual  
137 temperature for the site is 11 °C (January: 4 °C; July: 20 °C) and average annual precipitation is  
138 991 mm (PRISM Climate Group, 2020). The site is dominated by white fir, incense-cedar, and

139 sugar pine (*Pinus lambertiana*, Dougl.). No recent timber harvests have been conducted at the  
140 site. However, prescribed fire was applied across the study area in 2011 and 2012.

141 The Yosemite Pine (YOPI) site is located along Wawona Rd. 11 km from the southern  
142 entrance to Yosemite National Park in Mariposa County, CA. The site spans 397 ha and is  
143 managed by the National Park Service. Mean elevation within the site is 1722 m. Average annual  
144 temperature is 10 °C (January: 3 °C; July: 19 °C) and average annual precipitation is 1042 mm  
145 (PRISM Climate Group, 2020). Ponderosa pine (*Pinus ponderosa* Dougl. Ex Laws), and incense-  
146 cedar dominate the site. No recent timber harvests or fires have affected the study site.

147

### 148 2.1.2 Site Description for Snag Fall Rate Modeling Plots

149 From 1982 to 2001 researchers from the US Geological Survey installed a network of  
150 permanent plots ranging from 0.9 ha to 2.5 ha to study forest demography and dynamics in  
151 the old-growth mixed conifer forests of Sequoia-Kings Canyon and Yosemite National  
152 Parks. For this study, we use a subset of the data in the demography study. The study areas  
153 have never been logged and experienced frequent low to moderate severity surface fires prior to  
154 Euro-American settlement (Caprio & Swetnam, 1993). Four of our 23 demography plots and  
155 portions of the YOMI and SEKI study sites experienced prescribed burns or wildfires in  
156 recent decades. The demography plots encompass common mixed-conifer species in the  
157 southern Sierra Nevada, such as ponderosa pine, sugar pine, white fir, incense-cedar, Jeffrey  
158 pine, red fir (*Abies magnifica*, Pitcher), and black oak (*Quercus kelloggii*, Newb.). The  
159 elevations of plots range from 1500 to 2576 m. A portion of the demography plots in  
160 Sequoia-Kings Canyon and Yosemite National Parks overlap with the SEKI and YOMI fuel  
161 study areas, respectively.

162 *2.2 Field Measurements*

163 *2.2.1 Fuels*

164 In 2017, a collection of 29, 35, and 37 circular 0.05-hectare fixed radius plots were  
165 established at SEKI, YOMI, and YOPI, respectively. These plots were located randomly using a  
166 spatially stratified design (Stevens & Olsen, 2003). When establishing plots in 2017, crews  
167 monumented plot centers with a painted rebar stake for later remeasurements. In 2021, field  
168 crews located and re-measured all plots established in 2017. Crews recorded the diameter at  
169 breast height (height = 1.37m , DBH), species, height, and status (live or dead) of all standing  
170 dead trees with a DBH equal to or greater than 10 cm in 2017. Crews also recorded the decay  
171 status of snags in 2017, which classifies snag decay based on the condition of the tree's top,  
172 branches, twigs, sapwood, and heartwood (USFS, 2010). At each plot fuels were recorded along  
173 three transects with azimuths of 90, 210, and 330 degrees following the Brown's fuel protocol  
174 (Brown, 1974). Crews tallied 1-hour (<0.64 cm diameter) and 10-hour (0.64-2.54 cm) fuels from  
175 3 to 5 meters along each transect, 100-hour (2.54-7.62 cm) fuels from 3 to 6 meters along each  
176 transect. Depths of litter and duff were measured at 3 and 10 m along each transect. Crews  
177 measured the diameter and decay status (sound or rotten) of all 1000-hour (>7.62 cm) fuels along  
178 the entire length (12.62 m) of each transect. Fuels were measured in both 2017 and 2021.

179

180 *2.2.2 Snag Fall Rate Data*

181 In the Forest Demography plots (described in 2.1.2 above), all trees taller than 1.37 m in  
182 height were tagged, identified to species, measured for diameter, and the location of every tree  
183 was mapped. Crews visited plots annually to document tree mortality and status of snags and to  
184 tag and measure new (ingrowth) trees. Prior to 2013, crews did not explicitly document when

185 snags fell over, however, they noted when snags were not found or were found fallen onto the  
186 ground. In 2013, researchers combed through the annual plot notes to estimate the year snags fell  
187 (Battles. et al., 2015). Beginning in 2013, field crews began explicitly recording the year when  
188 snags fell. Snags were considered “fallen” if their main stem, or bole, broke at a height less than  
189 or equal to 1.37 m off the ground. We limited our analysis to snags greater than 12.7 cm in DBH.

190

## 191 *2.3 Data Analysis*

### 192 *2.3.1 Fuels*

193 We used the RFuels package (Foster et al., 2018) to calculate fuel loads for each fuel  
194 category (1-hour, 10-hour, 100-hour, 1000-hour, litter, duff). Transect level estimates were  
195 averaged to the plot level, then we used a paired t-test to determine if there were significant  
196 differences in fuel loads for each site from 2017 to 2021. The distributions of fuel loads were  
197 visually inspected to ensure the assumption of normality for the paired t-tests was met for all fuel  
198 categories. Prior to testing for significant differences, we aggregated the fuel calculations into  
199 four fuels categories: total fuel loads, fine woody debris fuel loads (1, 10, 100 hour), coarse  
200 woody debris fuel loads (1000-hour fuels), and litter and duff fuel loads added together into one  
201 category.

202

### 203 *2.3.2 Snag Fall Rates*

204 To evaluate snag fall rates, we applied survival analysis that uses the time of occurrence  
205 of an event to estimate the probability that the event has occurred by a specific time (Whitlock &  
206 Schluter, 2020). A key advantage of survival analysis is that not all subjects in a study need to

207 experience the event of interest within the time frame of the study. For this study, the event of  
208 interest was snag fall, and we were interested in assessing how long snags stood until they fall.

209         Using the demography dataset, we were able to determine the time each snag was  
210 standing. Many of the snags in our dataset had not fallen by the time of the most recent sample  
211 and were therefore treated as right-censored.

212         Some of the demography plots had experienced fire in recent decades. Trees killed by fire  
213 often fall at faster rates relative to snags created by other causes (Morrison & Raphael, 1993).  
214 Additionally, snags exposed to fire typically experience heightened fall rates compared to  
215 unburned snags (Landram et al., 2002). Since we were interested in understanding the fall rates  
216 of trees that died from non-fire related causes, we omitted snags created by fire and right-  
217 censored pre-existing snags that were exposed to fire. Specifically, we excluded any snags that  
218 were created within five years of a fire in our dataset. Research in the Sierra Nevada mixed  
219 conifer forests suggest that post-fire mortality levels off by roughly three years post-fire (Hood et  
220 al., 2010; Stephens & Finney, 2002). We excluded trees that died up to five years after fire to  
221 ensure that snags created afterward likely died from non-fire related causes and therefore would  
222 not bias our fall rate estimates. The right-censoring of snags allowed us to extract information  
223 about snag persistence from the time those snags were standing prior to the fire, while avoiding  
224 the possible confounding effect of fire on the survival rate post-fire.

225         We quantified the observed snag fall rate for each species using the Kaplan-Meier  
226 estimator, a non-parametric technique for estimating and visualizing survival probability over  
227 time (Kaplan & Meier, 1958). We also tested for differences in fall rates based on species using  
228 the log rank test, which tests the null hypothesis that survival curves for two or more groups are  
229 the same (Andersen & Keiding, 2014). We then used accelerated failure time (AFT) models to

230 project snag survival time. Conceptually, the AFT is a parametric model that defines a baseline  
 231 survival curve that can include covariates that speed up or slow down the survival times of  
 232 individuals (James, 2014). We evaluated the fit of five potential underlying distributions that are  
 233 commonly used when modeling snag fall (Parish et al., 2010; Grayson et al., 2022) rates:  
 234 Weibull, exponential, lognormal, loglogistic, and gaussian distributions. Based on Akaike  
 235 Information Criterion (AIC), the loglogistic distribution provided the best underlying fit for the  
 236 survival curve.

237 The survival function for the loglogistic distribution is defined by:

$$238 \quad S(t) = [ 1 + t^{1/\sigma} \exp(\frac{-\mu - \alpha_i x_i}{\sigma}) ]^{-1} \quad \text{Equation 1}$$

239 And the  $p$ th percentile of the survival distribution for the  $i$ th individual is  $t_i(p)$  defined as:

$$240 \quad t(p) = \exp [ \sigma \log(\frac{100-p}{100}) + \mu + \alpha_i x_i ] \quad \text{Equation 2}$$

241 Where:

$$242 \quad S(t) = \textit{probability a snag survives to a given time, } t$$

$$243 \quad \sigma = \textit{scale parameter}$$

$$244 \quad t = \textit{time}$$

$$245 \quad \mu = \textit{intercept}$$

$$246 \quad \alpha_i x_i = (\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_j x_j) \textit{ where } \alpha = \textit{covariate coefficients and } x = \textit{covariates}$$

247

248 We then fit multiple AFT models describing snag fall rates using snag DBH, snag  
 249 species, and the interaction of snag DBH and species as covariates. Due to small sample size and  
 250 the difficulty of field crews distinguishing between ponderosa pine and Jeffrey pine in the field,  
 251 we combined snags of these species into one species category of yellow pine to represent both  
 252 species. We selected the best model based on lowest AIC score. We assessed the fit of the best

253 model by comparing binned predicted survival probabilities to observed survival probabilities  
254 (Das et al., 2022). To do this, we predicted survival probabilities of each tree using our model,  
255 binned results in ten evenly divided classes (i.e., 0.1,0.2,0.3, up to 1.0), and then aggregated  
256 adjacent bins until there were at least 50 trees in each bin. We then compared the average  
257 predicted survival probability within each bin to the average observed survival probability within  
258 each bin. We also assessed model performance by calculating the concordance statistic, which is  
259 a measure of how well a model discriminates between dying and surviving objects (where a  
260 value of 1 indicates perfect discrimination and a value of 0.5 indicates discrimination that is no  
261 better than random). We summarized the effect size of species and DBH by generating predicted  
262 survival curves using the best fitting AFT.

263

### 264 *2.3.3 Snag fall rate and biomass*

265       Upon determining the best fitting model to describe snag fall rates, we used the 2017  
266 inventory of snags at SEKI, YOMI, and YOPI to forecast the input of biomass to the surface  
267 from falling snags at each site. We used regional biomass equations (Forest Inventory and  
268 Analysis 2010) to estimate live tree biomass that was originally represented by each snag when it  
269 was alive. We then discounted bole biomass in each snag using ratios based on decay class  
270 (Cousins et al.2015). Branch biomass in snags was also reduced based on decay class with no  
271 degradation in recently dead trees (decay class 1), 0.9 for young snags (decay class 2); 0.4 for  
272 middle-aged snags (decay classes 3 and 4), and 0 for old snags (decay class 5). The AFT allowed  
273 us to predict the year each snag will fall. Knowing the amount of biomass and expected fall date  
274 for each snag, we forecasted snag input to fuel loads at each site.

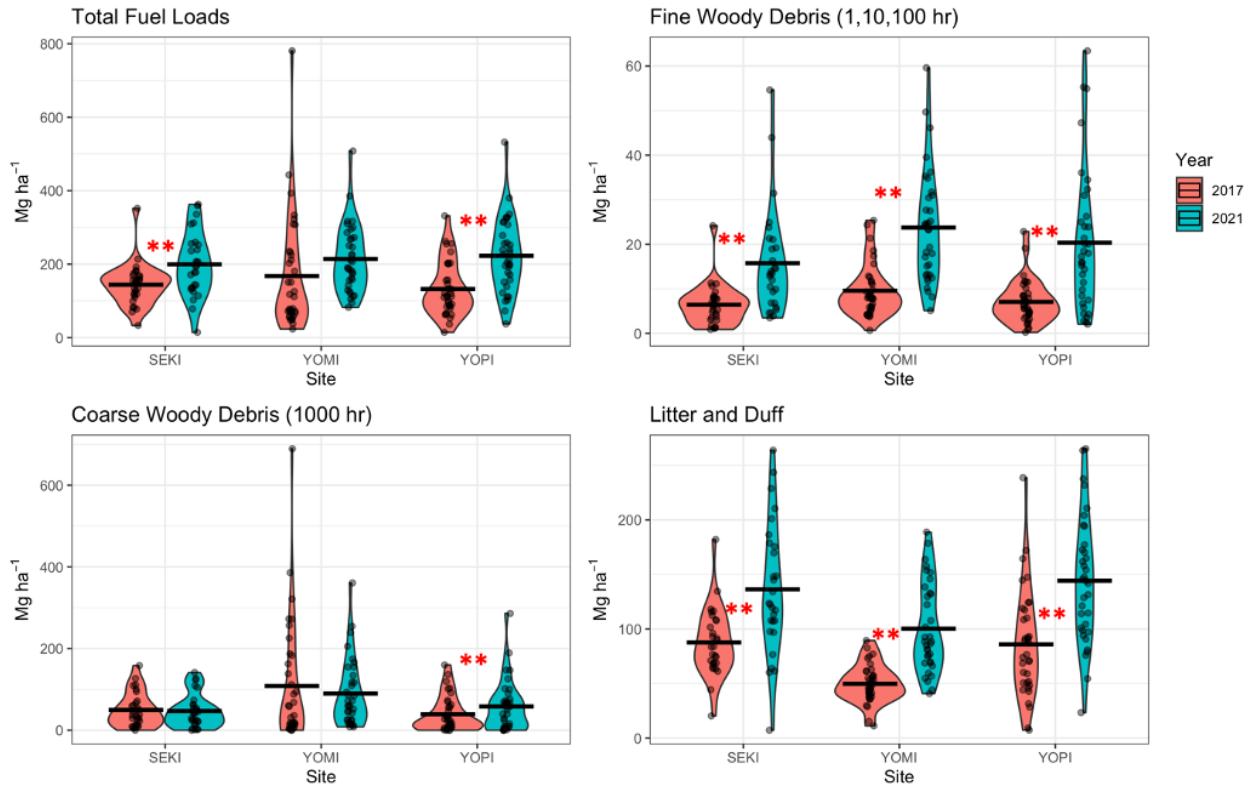
275 To characterize the variability and uncertainty associated with our model predictions, we  
276 generated forecasts using fall rates in 3 different percentiles, 25<sup>th</sup>, 50<sup>th</sup> (median), and 75<sup>th</sup>  
277 percentiles, which correspond to faster or slower overall fall rates. All analyses were conducted  
278 in the R statistical software (R Core Team, 2019) and survival modeling utilized the “survival”  
279 package in R (Therneau, 2015).

280

### 281 **3. Results**

#### 282 *3.1 Surface Fuel Load changes*

283 Total fuel loads increased significantly from 2017 to 2021 in SEKI and YOPI with  
284 increases of 55.43 Mg ha<sup>-1</sup> (38.5% increase) and 90.69 Mg ha<sup>-1</sup> (68.7%) ( $p < 0.05$ ,  
285 Supplementary Table S2, Figure 1), respectively but there was no change at YOMI ( $p = 0.136$ ).  
286 Fine woody debris fuel loads increased significantly at all 3 sites with average increases of 9.32  
287 (145.2%), 13.32 (189.2%), and 14.17 (148.1%) Mg ha<sup>-1</sup>, respectively ( $p < 0.05$ ). Coarse woody  
288 debris loads did not change significantly at SEKI and YOMI ( $p = 0.185$ ,  $p = 0.509$ ) while loads  
289 of coarse woody debris increased significantly by 18.96 Mg ha<sup>-1</sup> (48.6%) in YOPI ( $p = 0.032$ ).  
290 Litter plus duff loads increased significantly at all sites with increases of 48.58 (55.3%), 58.43  
291 (68%), and 50.54 Mg ha<sup>-1</sup> (101.5%) for SEKI, YOPI, and YOMI, respectively ( $p < 0.05$ ).



292

293 **Figure 1.** Violin plots of the fuel load estimates within each stratum at SEKI, YOMI, and YOPI  
 294 measured in 2017 (red) and 2021 (blue). Horizontal lines represent mean fuel loads for given site  
 295 and year. The colored region around each violin plot illustrates the proportion of data located at  
 296 that given value and the dots represent actual measurements for a given plot. Groups of plots  
 297 marked by \*\* denote a significant change from 2017 to 2021.

298

### 299 3.2 Snag Fall Rates

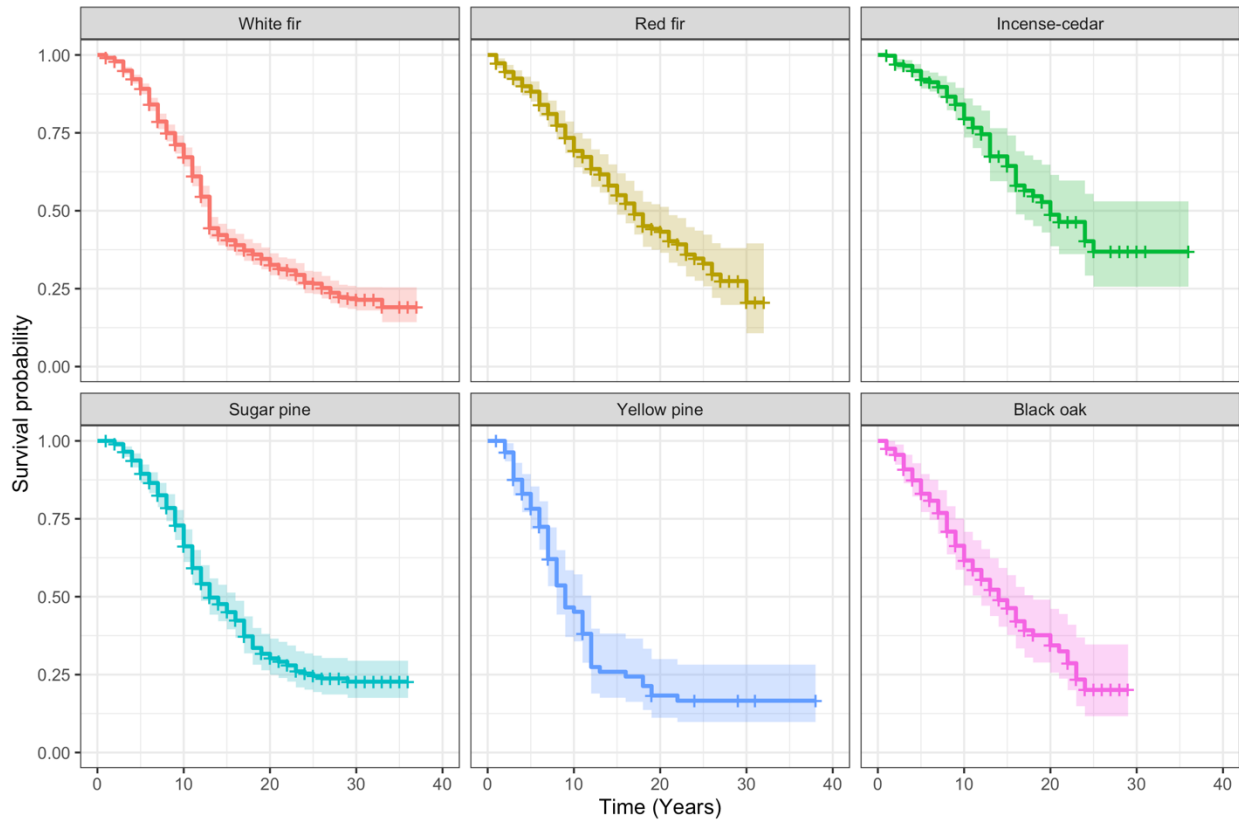
300 The USGS dataset observed 3,059 snags and 1,187 snag falls. Of the remaining 1,872  
 301 snags, 372 were right-censored due to fire and 1,500 did not fall during the observation period  
 302 (Table 1). DBH of snags ranged from 12.7 to 251 cm. Diameter distributions for each species  
 303 followed the general reverse J-shaped curve (Supplementary Figure S1).

Species	Number of snags	Number of snags that fell	Median Survival Time (25 <sup>th</sup> – 75 <sup>th</sup> percentile) (Years)	Mean DBH (cm) (Range)
White fir	1418	604	13 (8-26)	43 (12.7-162)
Red fir	412	147	17(9-27)	49.4 (12.7 – 251)
Incense-cedar	412	71	20 (13-UND)*	25.2 (12.7-184)
Sugar pine	483	211	13 (9-25)	41.1 (12.7-98.1)
Yellow pine	175	76	9 (6-16)	49.9 (12.7-162)
Black oak	159	78	14 (8-23)	23.6 (12.7-98.1)

304 **Table 1.** List of species included in the model, number of snags of each species, number of snags  
305 that fell for each species, median survival time using Kaplan-Meier estimator, and mean DBH.

306 \*75<sup>th</sup> percentile is undefined due to skewedness of data.

307



308

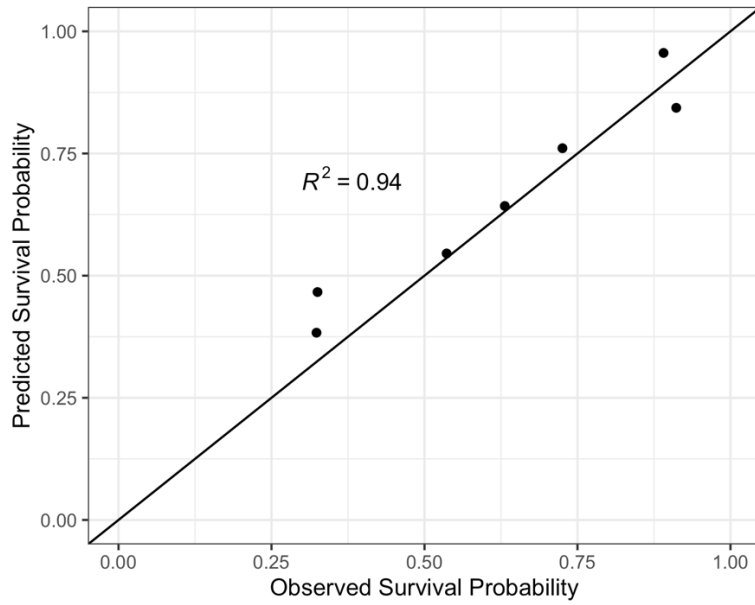
309 **Figure 2.** Observed survival curves using Kaplan-Meier estimator for each species with 95%  
 310 confidence intervals (shaded areas). Curves depict survival probability of snags over time, by  
 311 species.

312

313 Median survival times (Table 1) (Kaplan-Meier estimator), and the results from the log-  
 314 rank test using species as groups ( $\chi^2 = 47.8, df = 5, p < 0.01$ ) indicate that snag fall rates  
 315 vary by species. Yellow pine snags had the shortest survival time with a median of 9 years while  
 316 incense-cedar had the longest survival time with a median of 20 years.

317 The best performing AFT model used a log-logistic distribution and included the  
 318 interaction of species and diameter as covariates. This modeled yielded the lowest AIC score  
 319 (Supplementary Table S4), demonstrated good model performance with a concordance statistic  
 320 of 0.6951 (SE = 0.009) and revealed a strong fit to the data (Figure 3).

321



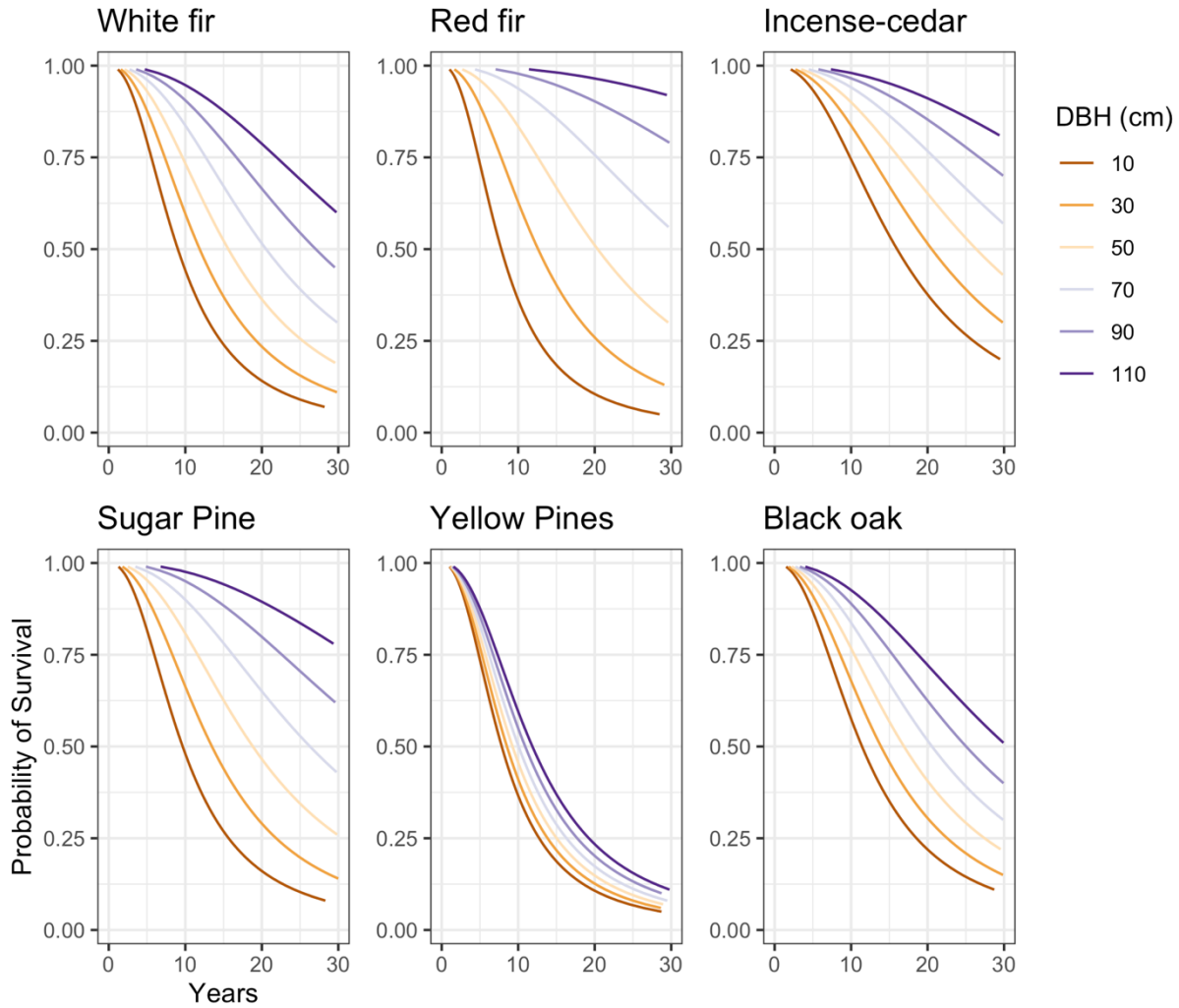
322

323 **Figure 3.** Average binned predicted survival probabilities calculated for all trees used to fit the  
324 top model ( $n = 3059$ ), compared to actual proportion of trees that fell within a bin. Pseudo  $R^2$   
325 value reported on figure.

326

327

328



329

330 **Figure 4.** Predicted mixed conifer survival times for each species of different sizes.

331

332 Model results reveal a strong effect of both species and size and they highlight key  
 333 interactions between species and size (Figure 4, Supplementary Table S3 and Figure S2). Across  
 334 all species snag longevity increased with size, however, the effect varied by species. Size played  
 335 a small role in influencing yellow pine fall rates but a very large role in red fir and sugar pine  
 336 longevity. Size played an important, but less pronounced role for white fir, incense-cedar, and

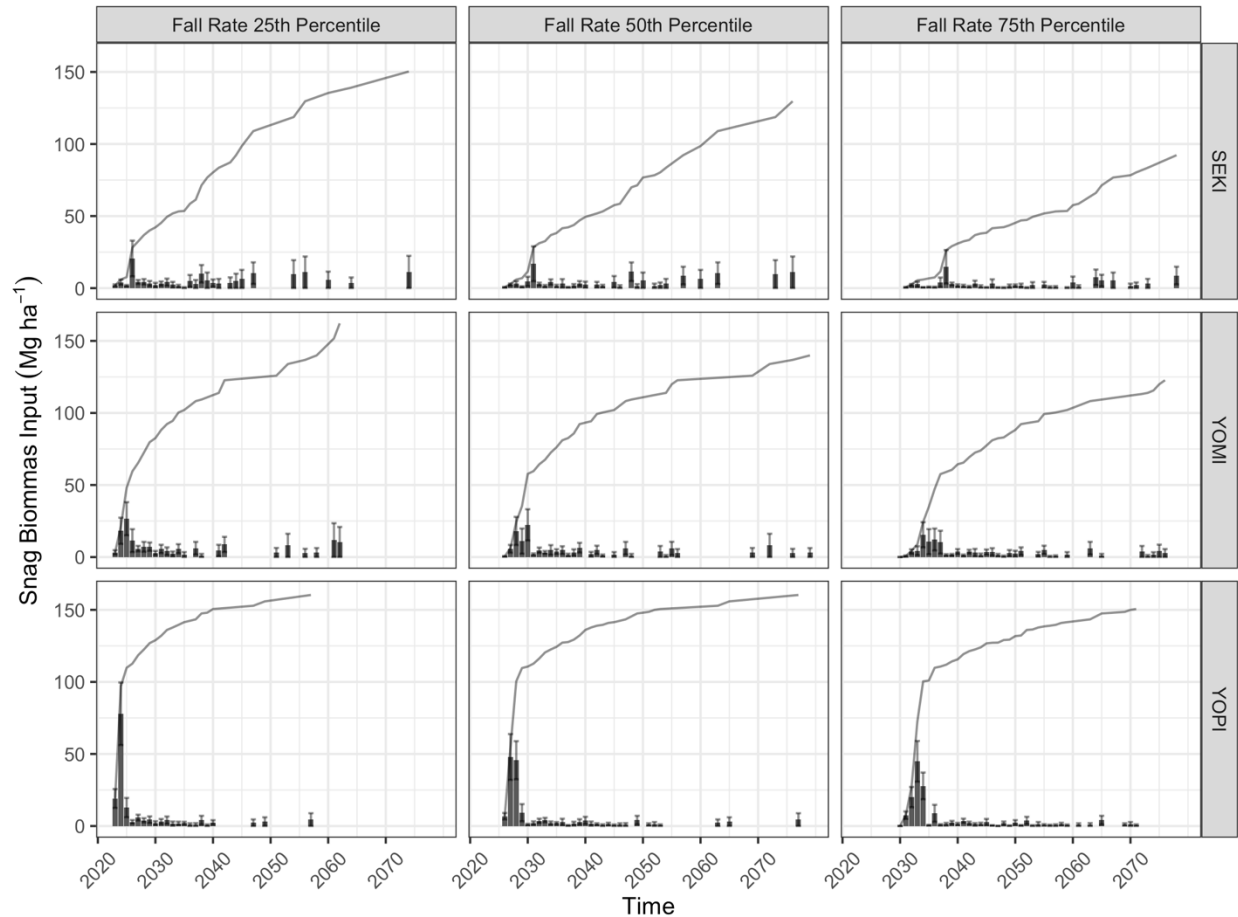
337 black oak. When size is held constant, yellow pines fall the fastest while incense-cedars fall the  
338 slowest.

339

### 340 *3.3 Snag Fall Rates and Fuel Loads*

341 Our projections suggest there will be large inputs of biomass from snag fall in the coming  
342 decades in these Sierra Nevada mixed conifer forests. Based on 2017 field measurements, SEKI,  
343 YOMI, and YOPI have an average of 162.4 (SE=31.07), 236.8 (64.05), and 165.08 (24.39) Mg  
344 ha<sup>-1</sup> of standing dead biomass in the form of snags, respectively. Our projections show that the  
345 rate of input from snag fall varies across our three sites with YOPI having the fastest input of  
346 snag material, with projections of 136.1 Mg ha<sup>-1</sup> (115.6-150.6, 25<sup>th</sup> and 75<sup>th</sup> percentile values) of  
347 biomass will fall by 2040 (Figure 5). YOMI is predicted to have a fast initial deposition of  
348 material, with 92.3 Mg ha<sup>-1</sup> (64.4-109.3) predicted to fall by 2040 followed by a slower but  
349 steady input for the decades afterward (Figure 5). SEKI is projected to have the slowest biomass  
350 deposition with 49.4 Mg ha<sup>-1</sup> (30.9-80.3) predicted to fall by 2040, and the cumulative biomass  
351 curve revealing a somewhat linear cumulative input over time (Figure 5).

352



353

354 **Figure 5.** Forecasted biomass inputs over time based on current standing snags and model  
 355 predictions at SEKI, YOPI, and YOMI, based on 25<sup>th</sup>, 50<sup>th</sup>, and 75<sup>th</sup> percentile modeled fall rates.  
 356 Bars depict forecasted annual biomass input from snag fall, the error bars depict standard error of  
 357 annual biomass input, and the lines depict cumulative biomass input from snag fall.

358

#### 359 **4. Discussion**

##### 360 *4.1 Short- term changes in fuel loads*

361 Across all 3 sites we measured a significant increase in the amount of fine woody  
 362 material and litter and duff from 2017-2021. These results align well with theoretical timelines of  
 363 fuel loads following bark beetle mortality (Hicke et al., 2012; Stephens et al., 2018) as well as

364 empirical observations of fuel loads following similar mortality events in Rocky mountain  
365 lodgepole pine (*Pinus contorta* var. *latifolia* Englem. Ex Wats) forests (Klutsch et al., 2009; Page  
366 & Jenkins, 2007; Schoennagel et al., 2012), all of which propose or observe heightened fine  
367 woody material and litter loads as recently killed trees quickly lose leaves and fine twigs.  
368 Raphael and Morrison (1987) found that pine and fir snags in the Sierra Nevada lost most of  
369 their foliage and small branches within 5 years of mortality which suggests that the decay of  
370 snags created by the 2012-2016 drought are likely the primary source of the increases we  
371 observed.

372         There was only a significant increase in coarse woody debris loads at the YOPI site  
373 whose overstory is predominantly ponderosa pine (Supplementary Table S1). This confirms the  
374 implications from our fall rate model (Figure 4) and fuel projections (Figure 5) which show  
375 faster fall rates in yellow pines and the fastest input of biomass from snag fall at the YOPI site.

376

#### 377 *4.2 Snag fall rates*

378         Our observed fall rates generally align with what has been reported in the past regarding  
379 mixed conifer species. Many studies report that ponderosa pine snags fall faster than fir snags  
380 and that incense-cedar snags often fall at the slowest rates (Grayson et al., 2022; Raphael &  
381 Morrison, 1987; Ritchie et al., 2013) which is consistent with our findings. While our  
382 observation that snag longevity increases with DBH is also common (Keen, 1955; Parish et al.,  
383 2010), an interactive effect between species and size has not been reported. Larger trees tend to  
384 have a greater volume of rot-resistant heartwood which provides structural support to standing  
385 trees (Harmon et al., 1986), and would explain the general trend that larger snags remain  
386 standing longer relative to smaller snags. The interaction of size and species is likely driven by

387 varying amounts of sapwood and heartwood within each species and the processes by which  
388 each species decays. Yellow pines trees decay much more rapidly than other species (Raphael &  
389 Morrison, 1987) so perhaps size is less important for these species since extra stability-providing  
390 heartwood simply degrades quickly, meaning larger trees only provide marginally more stability  
391 and snag survival time, as suggested in Figure 4. Ultimately, snag failure is most often a result of  
392 decay that has removed structural stability. The most common agents of wood decay are insects  
393 and fungi which often have species-specific interactions with snags (California Forest Insect and  
394 Disease Training Manual, 2006). Furthermore, rate of bark loss, aspect, and elevation also play a  
395 large role in decay (Audley et al., 2021; Harmon et al., 1986; Kimmey, 1955) so variation in  
396 those qualities likely contributed to the observed differences in the effects of species and size on  
397 fall rates.

398

### 399 *4.3 Projection Long Term Changes in Fuel Loads*

400 Our 3 study sites are projected to have large inputs of coarse woody material in the  
401 coming years as a result of the 2012-2016 drought (Figure 5). The YOPI site is projected to have  
402 the fastest inputs which is driven by the fast fall rates of yellow pine snags, whereas the SEKI  
403 and YOMI sites have more gradual inputs over time, likely due to a more mixed snag  
404 composition that yields varying and longer snag survival times. Regions of the Sierra Nevada  
405 that contain high densities of yellow pine snags will likely see rapid fall rates and heavy coarse  
406 woody material surface fuels more quickly similar to our projections for YOPI. On the contrary,  
407 regions with more balanced or less pine dominated snag compositions may see more gradual  
408 deposition of snag material, similar to YOMI and SEKI.

409

#### 410 *4.4 Model and Projection Limitations*

411           There are a few assumptions and limitations to our modeling and projections that warrant  
412 discussion. A key assumption is that we assumed trees dead in 2017 would contain the same  
413 amount of biomass by the time of their predicted fall date. In reality, these trees will have  
414 decayed and lost some of their biomass over time, as reported in Cousins et al. (2015), implying  
415 that the actual biomass inputs from snags may be lower than those reported here. Our forecast  
416 also only considers snags present in 2017, and therefore omits fuel added from trees that die in  
417 the future.

418           Our model and projections also overlook other factors that are known to influence snag  
419 fall rates, so extrapolating our projections to other regions should be done with caution. For  
420 instance steeper slopes, moist soils, and southern aspects are known to increase fall rates while  
421 higher elevations, northern aspects, and higher canopy cover have been associated with lower  
422 fall rates (Audley et al., 2021; Keen, 1955; Rhoades et al., 2020; Ritchie et al., 2013). Since there  
423 is certainly variability in these conditions, it follows that the magnitude and timing of biomass  
424 inputs may vary from our predictions. Nonetheless, our results still highlight that a large amount  
425 of biomass is expected to be input to the surface within the next 20 years in the Sierra Nevada.

426           Snag persistence is a difficult process to model because there are many stochastic  
427 processes that can dramatically alter fall rates that are difficult to capture and include in a  
428 predictive model. Despite these difficulties, our top model revealed good performance (Figure 3)  
429 and had a high concordance statistic of 0.6951 which is higher than values reported in other snag  
430 fall rate model studies (Audley et al., 2021), and is within the range of values reported in Das et  
431 al. (2021). We emphasize that the large sample size (n=3059), annual resolution, and long  
432 temporal extent (up to 39 years) of our data contributed to the performance of our model.

#### 433 *4.5 Management Implications and Conclusion*

434           Within our study sites and likely across a great deal of the western Sierra Nevada that  
435 experienced drought-related mortality, fine fuel loads have already increased, and coarse woody  
436 debris loads will rise in the coming 10-20 years as many snags fall to the ground. This increase  
437 in fuel loads from snag-fall, coupled with already heavy surface fuel loads, will increase the risk  
438 of high severity fire and potential for mass fire (Arno, 2000; Brown et al., 2003; Coppoletta et  
439 al., 2016; Stephens et al., 2018). There will be increased difficulties for wildland firefighting  
440 operations as snags falling represent a considerable danger to field crews and downed snags can  
441 hinder fireline construction and block vehicle access (Dunn et al., 2019; Plucinski, 2019).  
442 Managers should anticipate particularly fast fall rates and high coarse woody debris loads in  
443 areas containing high densities of large yellow pine snags, which were one of the most common  
444 targets of the recent bark beetle and drought mortality (Axelson et al., 2019; Fettig et al., 2019;  
445 Pile et al., 2019). We also highlight that the surface fuel loads we observed in 2017 prior to  
446 increases associated with the drought mortality were much greater than estimated historical loads  
447 in the Sierra Nevada or reported in the Sierra San Pedro Martir, a contemporary reference forest  
448 that never experienced logging or extensive fire suppression until the 1980s (Safford & Stevens,  
449 2017; Stephens, 2004; Stephens et al., 2007).

450           Managers are not without options for dealing with the heightened surface fuels and large  
451 ground combustible biomass associated with extreme snag densities. Salvage logging operations  
452 could recover some of the economic value in the snags while also removing some of the snag  
453 biomass to lower future fuel input from snags (Prestemon et al., 2006). This is sometimes limited  
454 due to the rapid decline in timber value following tree mortality and the large amount of area  
455 inaccessible to logging operations in the Sierra Nevada due to factors such as protected habitats

456 and rugged topography, particularly on federally-owned land (North et al., 2015). Salvage  
457 operations may still be an option for some areas given that much of the Sierra Nevada is  
458 privately-owned and subject to fewer habitat restrictions (Stewart et al., 2016). Although  
459 expensive, non-economic salvage operations could also be conducted such as falling, piling, and  
460 burning snags to reduce snag biomass and future fuel inputs of coarse woody debris. While  
461 salvage operations can lower future inputs of fuel from snag fall, they cannot reduce surface  
462 fuels already present on the ground which is a key characteristic of successful forest restoration  
463 (Stephens et al., 2020).

464 To address this concern managers may consider using prescribed burning in areas of high  
465 mortality to consume recently deposited fine fuels from mortality and any coarse woody debris  
466 already present from snag fall (Knapp et al., 2005; Stephens et al., 2018). Our results suggest  
467 most snags will fall within the next 10-20 years (Figure 5). Since operating in areas of high snag  
468 densities can be dangerous it would be best to burn within the next few years before the period of  
469 frequent snag fall or to burn after most snags have fallen in 20 or more years to avoid burning  
470 under high-risk snag situations. It may be best to burn prior to snag fall to remove the high  
471 surface fuel loads already on the ground so that future coarse woody debris from snag fall is not  
472 embedded within already heightened surface fuels that increase risk of mass fires (Stephens et  
473 al., 2018). Burning is also likely to change the rate of snag fall and fuel input since snags  
474 exposed to fire often experience heightened fall rates afterward (Landram et al., 2002). Managers  
475 may be able to reduce fuels most-efficiently by burning prior to the snag fall and then reburning  
476 10-20 years later after most snags have fallen to consume recently downed material as well. It is  
477 important to acknowledge the uncertainty surrounding these treatment options given that our  
478 projections do not incorporate the influence of fire on snag fall rates, burn operations may be

479 difficult in stands with high log densities, and coarse woody debris consumption may be limited  
480 under prescribed burning conditions (Knapp et al., 2005). Despite this uncertainty and the many  
481 regulatory, operational, and political barriers to implementing prescribed burning (Miller et al.,  
482 2020; York et al., 2020) it will remain a valuable tool for managers to consume the large  
483 amounts of combustible material associated with the 2012-2016 drought mortality event.

484 For forest managers, understanding the longevity of snags is crucial for future planning  
485 and decision-making, regarding minimizing risk for recreation and firefighter access to forests. A  
486 common tool used by managers to approach this issue is the Fire and Fuels Extension of the  
487 Forest Vegetation Simulator (FVS) which includes submodels that describe snag dynamics  
488 (Reinhardt & Crookston, 2003). Recent work by Grayson et al. (2019) suggests that FVS  
489 inaccurately predicts postfire snag fall rates and needs revision. We performed a comparison of  
490 predicted survival times using FVS, equations reported by Grayson et al. (2019), and our top  
491 performing model. Our comparison revealed strong difference in predictions with Grayson et al.  
492 (2019) reporting faster fall rates and FVS predicting slower fall rates (Table 2). We believe this  
493 highlights not only the importance of revising fall equations used in FVS but also the importance  
494 in using different fall rate equations based on the cause of death for snags since strong difference  
495 in fall rates for fire-killed versus non fire-killed snags have been reported (Grayson et al., 2019,  
496 this work).

497

498

499

500

Prediction Tool	Median predicted survival time (years)	Number of snags predicted to fall within 10 years	Number of snags predicted to fall within 20 years
Our top model	12.6	238	664
Grayson et al. 2019 equations	10.31	307	655
Forest Vegetation Simulator	14.75	0	556

501 **Table 2.** Comparison of predicted snag fall rates for snags from 2017 dead tree survey (771  
502 snags), using our top model, equations reported in Grayson et al. (2019) and using predictions  
503 from the Forest Vegetation Simulator tool (Reinhardt & Crookston, 2003).

504

505         There are many difficulties and risks associated with managing forests following  
506 extensive mortality and providing managers with better tools and information to predict and  
507 anticipate snag fall rates will help address some of these challenges. This study provides  
508 managers with some essential information regarding snag longevity and fuel load accumulation  
509 following the extensive drought in the Sierra Nevada that can aid in future decision making.

510

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746 **Supplementary Materials**

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Site	Mean Live Density (SE) (trees ha <sup>-1</sup> )	Mean Dead Density (SE) (trees ha <sup>-1</sup> )	Mean Live Basal Area (SE) (m <sup>2</sup> ha <sup>-1</sup> )	Mean Dead Basal Area (SE) (m <sup>2</sup> ha <sup>-1</sup> )	Dominance (%)
SEKI	312.7 (33.4)	121.8 (22.2)	41.9 (4.8)	21.4(3.3)	White fir (32%) Incense-cedar (26%)
YOMI	201.5(17.3)	147.4 (17.9)	51.2 (5.9)	26.2 (5.0)	Incense-cedar (25%) White fir (20%)
YOPI	351.7 (32.8)	205.6 (31.9)	39.6 (6.4)	32.2 (5.4)	Ponderosa pine (55%) Incense-cedar (28%)

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749 **Supplementary Table S1.** Overstory structure and composition from 2017 tree inventory at sites  
 750 where fuel loads were measured. Dominance shows 2 most dominant tree species observed based  
 751 on relative basal area percentages at each site.

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Site	Fuel Strata	2017 (SE) Mg ha <sup>-1</sup>	2021 (SE) Mg ha <sup>-1</sup>	Mean Change (SE) Mg ha <sup>-1</sup>	t-value	P-value
<b>SEKI</b>	<b>Total</b>	<b>143.83</b> <b>(11.04)</b>	<b>199.26</b> <b>(16.28)</b>	<b>55.43</b> <b>(17.97)</b>	<b>3.0844</b>	<b>0.004667</b>
<b>YOPI</b>	<b>Total</b>	<b>132</b> <b>(12.42)</b>	<b>222.84</b> <b>(16.97)</b>	<b>90.69</b> <b>(13.63)</b>	<b>6.6562</b>	<b>1.226e-07</b>
YOMI	Total	167.63 (26.16)	214.05 (15.48)	46.42 (30.42)	1.5261	0.1362
<b>SEKI</b>	<b>FWD</b>	<b>6.42</b> <b>(0.88)</b>	<b>15.75</b> <b>(2.25)</b>	<b>9.32</b> <b>(2.26)</b>	<b>4.1275</b>	<b>0.0003156</b>
<b>YOPI</b>	<b>FWD</b>	<b>7.04</b> <b>(0.82)</b>	<b>20.36</b> <b>(2.69)</b>	<b>13.32</b> <b>(2.48)</b>	<b>5.3742</b>	<b>5.609e-06</b>
<b>YOMI</b>	<b>FWD</b>	<b>9.57</b> <b>(0.99)</b>	<b>23.73</b> <b>(2.11)</b>	<b>14.17</b> <b>(2.11)</b>	<b>6.7121</b>	<b>1.04e-07</b>
SEKI	CWD	49.64 (7.91)	47.17 (7.72)	-2.48 (10.16)	-0.24357	0.8094
<b>YOPI</b>	<b>CWD</b>	<b>39.21</b> <b>(7.27)</b>	<b>58.17</b> <b>(10.64)</b>	<b>18.96</b> <b>(8.49)</b>	<b>2.2326</b>	<b>0.03227</b>
YOMI	CWD	108.26 (24.96)	89.97 (13.93)	-18.29 (27.42)	-0.66696	0.5093
<b>SEKI</b>	<b>Litter and Duff</b>	<b>87.77</b> <b>(5.87)</b>	<b>136.35</b> <b>(11.34)</b>	<b>48.54</b> <b>(10.14)</b>	<b>4.7884</b>	<b>5.379e-05</b>
<b>YOPI</b>	<b>Litter and Duff</b>	<b>85.89</b> <b>(8.27)</b>	<b>144.32</b> <b>(9.76)</b>	<b>58.43</b> <b>(7.49)</b>	<b>7.79</b>	<b>4.51e-09</b>
<b>YOMI</b>	<b>Litter and Duff</b>	<b>49.8</b> <b>(2.94)</b>	<b>100.35</b> <b>(6.85)</b>	<b>50.54</b> <b>(7.74)</b>	<b>6.5266</b>	<b>1.798e-07</b>

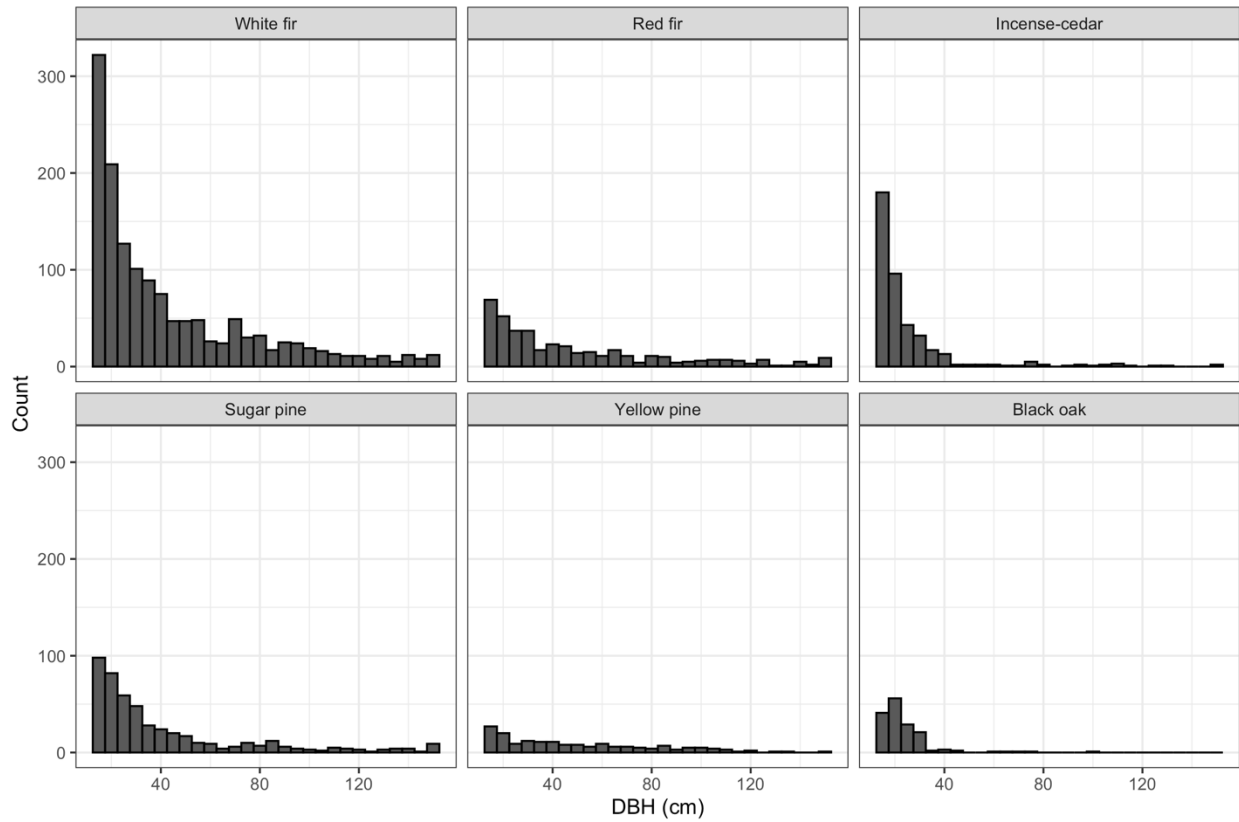
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762 **Supplementary Table S2.** Mean fuel loads in each fuel strata, at each site, in 2017 and 2021,

763 mean change in fuel loads for each stratum, and result of paired t-test between years at SEKI,

764 YOMI, and YOPI. Values in bold denote significant change at an alpha level of 0.05.

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767 **Supplementary Figure S1.** Diameter distributions for each species included in our analysis of

768 snag fall rates. All trees with a DBH greater than 150 cm were placed into a 150+cm bin to

769 improve visual appearance of histograms. Yellow pine includes ponderosa pine and Jeffrey pine.

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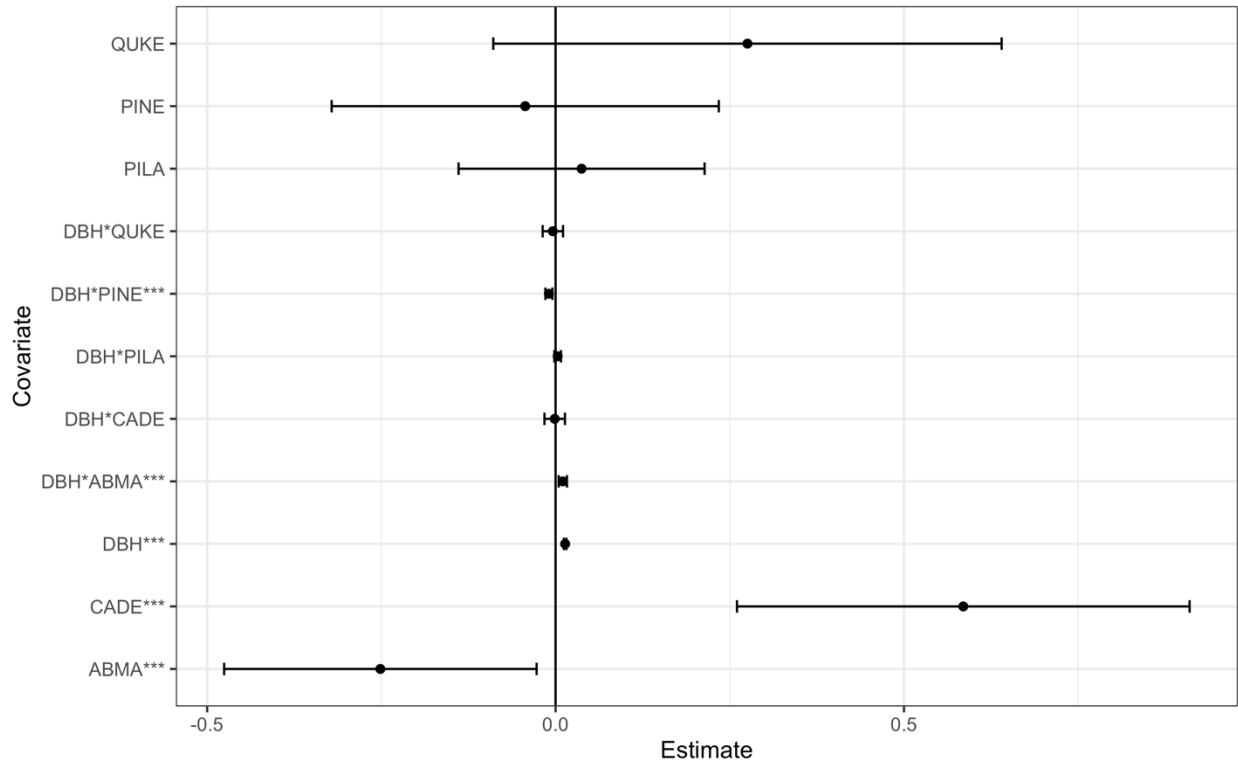
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Covariate	Estimate	Standard Error	Z	P-value
<b>Intercept</b>	<b>2.06138</b>	<b>0.042679</b>	<b>48.34</b>	<b>&lt;2e-16</b>
<b>DBH</b>	<b>0.013717</b>	<b>0.000943</b>	<b>14.55</b>	<b>&lt;2e-16</b>
ABCO	Reference	na	na	na
<b>ABMA</b>	<b>-0.251443</b>	<b>0.114400</b>	<b>-2.20</b>	<b>0.02795</b>
<b>CADE</b>	<b>0.585106</b>	<b>0.165697</b>	<b>3.53</b>	<b>0.00041</b>
PILA	0.037314	0.090075	0.41	0.67869
PINE	-0.043574	0.141739	-0.31	0.75852
QUKE	0.275379	0.186104	1.48	0.13895
DBH*ABCO	Reference	na	na	na
<b>DBH*ABMA</b>	<b>0.010341</b>	<b>0.003049</b>	<b>3.39</b>	<b>0.00069</b>
DBH*CADE	-0.001248	0.007489	-0.17	0.86764
DBH*PILA	0.002984	0.002431	1.23	0.21956
<b>DBH*PINE</b>	<b>-0.009595</b>	<b>0.002550</b>	<b>-3.76</b>	<b>0.00017</b>
DBH*QUKE	-0.003921	0.007460	-0.53	0.59922
<b>Log(scale)</b>	<b>-0.819829</b>	<b>0.022814</b>	<b>-35.94</b>	<b>&lt;2e-16</b>

778 **Supplementary Table S3.** Estimated parameters of the accelerated failure time model with  
779 lowest AIC predicting snag survival time. Values in bold denote significant change at an alpha  
780 level of 0.05.

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**Supplementary Figure S3.** Estimated parameters and 95% confidence intervals of the

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accelerated failure time model with lowest AIC predicting snag survival time. Parameter

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estimates that are significant at  $\alpha = 0.05$  are labeled with \*\*\*

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Model Covariates	AIC	$\Delta AIC$	Concordance (SE)
DBH * Species	9046.007	0	0.6951 (0.008929)
DBH + Species	9068.152	22.145	0.6846 (0.008895)
DBH	9177.373	131.366	0.6541 (0.009063)
Species	9454.527	408.52	0.552 (0.009739)
Null	9493.797	447.79	0.5 (0)

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**Supplementary Table S4.** Model covariates, AIC scores, and concordance statistics of all 5 log-

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logistic AFT models.