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**Title**

Trends in spatial patterns of stand-replacing fire in California mixed-conifer forests, 1984-2015

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**Abstract**

Stand-replacing fire has profound ecological impacts in drier mixed-conifer forests, yet there is continued uncertainty over how best to describe the scale of stand-replacing effects within individual fires, and how these effects are changing over time. In forests where regeneration following stand-replacing fire occurs via seed dispersal from surviving trees, the spatial scale and pattern of stand-replacing effects is a critical metric that is often overlooked. We utilized a novel, recently developed metric that describes the amount of stand-replacing area within a given distance of a stand-replacing patch edge, in order to compare fires that may be otherwise similar in the size of the fire or the percentage of stand-replacing effects. Specifically, we analyzed 477 fires in California mixed-conifer forests between 1984 and 2015 and asked whether this metric, the stand-replacing decay coefficient (SDC), has changed over time, whether it is affected by fire management and past forest management, and how it responds to extreme weather conditions at the time of the fire. SDC decreased over time, indicating that stand-replacing patches became larger and more regularly shaped. The decrease in SDC was particularly pronounced in the years since 2010. While SDC is correlated with percent high-severity, it is able to distinguish fires of comparable percent high-severity but different spatial pattern, with managed fires having higher SDC than suppressed fires, and fires managed by the National Park Service having higher SDC than fires managed by the US Forest Service. Fire weather also played an important role, with higher maximum temperatures generally associated with lower SDC values. SDC is a useful metric to compare fires in this way, because it is associated with more conventional metrics such as percent high-severity and is also directly associated with distance to surviving trees at the patch edge, the biological mechanism driving resilience following stand-replacing fire in mixed-conifer forests.

**Introduction**

In forests, tree mortality from fire is an important ecological process that promulgates changes in forest structure, fuel loads, vegetation diversity and wildlife habitat suitability ([Swanson et al. 2011](#_ENREF_38)). Tree mortality during a fire is a binary process (a tree is top-killed or not), but it is not spatially independent: weather, fuel or topographic conditions that lead to the mortality of one tree also increase the likelihood of mortality for neighboring trees (Finney?). When a patch of adjacent trees are all killed by fire, this is termed “stand-replacing fire”. This term is scale-independent – stand-replacing fire can refer to sub-ha stands of ≤100 trees, or to many-ha stands of >10,000 trees – but the implications of the spatial scale of stand-replacing fire are profound.

Forest resilience following stand-replacing fire depends on ecological memory in the form of tree propagules ([Johnstone et al. 2016](#_ENREF_14)). In forests where the dominant tree species have evolved the ability to propagate after being top-killed by fire, e.g. via basal resprouting in oaks (*Quercus spp*.) or serotinous cones in Rocky Mountain lodgepole pine (*Pinus contorta var. latifolia*), resilience is maintained even in large stand-replacing patches. In forests where the dominant tree species lack these adaptations (e.g. many western mixed-conifer forest types), propagules must arrive via surviving trees on the edges of stand-replacing patches, and the size and shape of these patches becomes critical. Thus forest resilience is reduced when contiguous stand-replacing patches become larger, tree regeneration towards the interior of these patches is slowed by dispersal limitation, and the likelihood of future stand-replacing fire within these patches increases ([Stevens et al. 2014](#_ENREF_37), [Coppoletta et al. 2016](#_ENREF_8), [Johnstone et al. 2016](#_ENREF_14), [Welch et al. 2016](#_ENREF_41)).

This potential for large-scale tree regeneration failure and persistent type-conversion, rather than negative effects of stand-replacing fire *per se*, is what drives much of the concern over stand-replacing fire in mixed-conifer forests ([Millar and Stephenson 2015](#_ENREF_20)). As such, there have been numerous attempts to quantify trends in the extent of stand-replacing fire in contemporary wildfires and infer how climate and forest management practices (e.g. historical fire suppression and firefighting tactics) might drive these trends ([Miller et al. 2009b](#_ENREF_25), [Miller and Safford 2012](#_ENREF_24), [Miller et al. 2012b](#_ENREF_26), [Harvey et al. 2016b](#_ENREF_12), [Picotte et al. 2016](#_ENREF_31)).

Most efforts to quantify trends in stand-replacing fire rely on interpretation of satellite-based vegetation change indices, particularly the differenced Normalized Burn Ratio (dNBR) and a version of that ratio relativized to pre-fire vegetation cover (RdNBR) ([Miller and Thode 2007](#_ENREF_27)). Burn severity (the amount of dominant vegetation killed or consumed by fire within a given area) can then be estimated by calibrating this ratio to field-derived data on canopy cover loss from fire, basal area loss from fire, or other composite field indices of burn intensity ([Miller et al. 2009a](#_ENREF_22)), generally at the scale of a 30-m LANDSAT pixel. Modern burn severity classifications transform a continuous variable (e.g. RdNBR) into a discrete variable at the pixel scale (e.g. “low”, “moderate” or “high” severity), based on threshold values of RdNBR associated with particular field conditions (e.g. ≤20%, 20-70%, or >70% basal area mortality). Field validations of post-fire mixed-conifer stands mapped as “high-severity”, whether using a 70% or a 90% basal area mortality threshold, indicate these areas generally have >>95% basal area mortality, with 100% basal area mortality being by far the most common condition greater than 30 m from the edge of a patch mapped as “high-severity” ([Miller and Quayle 2015](#_ENREF_23), [Lydersen et al. 2016](#_ENREF_18)). Thus, areas of “high-severity fire” mapped in this way are reasonable approximations of “stand-replacing fire”.

More recently, the term “mixed-severity fire” has become popular to describe individual fires, or characteristic effects of multiple fires (i.e. fire regimes), wherein some fraction of a burned area experiences stand-replacing effects. Because previous discrete classifications of fire effects defined low-, moderate- and high-severity fire regimes as experiencing intense fires ≤20%, 20-70%, or >70% percent of the time they burned ([Agee 1993](#_ENREF_2), [Agee 1998](#_ENREF_3)), “mixed-severity fires” are commonly defined as those wherein 20-70% of the fire experiences stand-replacing effects ([Perry et al. 2011](#_ENREF_30), [Hessburg et al. 2016](#_ENREF_13)). These fires are often identified as having 20-70% of their area mapped as high-severity using satellite-based classifications ([Perry et al. 2011](#_ENREF_30)), despite the fact that tree mortality is also present in areas of the fire mapped as low- and moderate-severity. This approach highlights the critically important role of spatial scale when classifying fire effects: tree mortality from fire is often spatially contiguous, and patches of stand-replacing fire of *ecologically meaningful size* are those mapped as “high-severity”. A fire mapped entirely as “moderate-severity” would by definition be a “mixed-severity” fire, with 20-70% basal area mortality at a very fine grain, yet fires where even a majority of area is mapped as “moderate-severity” are exceedingly rare ([Belote 2015](#_ENREF_4), [Harvey et al. 2016a](#_ENREF_11)). Mixed-severity fires generally produce discrete patches of stand-replacing fire, eventually filled in by grass, shrubs, or tree regeneration, surrounded by surviving forest that burned at low- to moderate-severity.

While the “patchy” nature of mixed-severity fires leads to a wide range of potential patch sizes and shapes, the conventional definition of mixed-severity fire says nothing about these attributes. Percent high-severity is a useful way to measure fire effects and compare among multiple fires, as it is easily derived and easily interpretable ([Miller et al. 2009b](#_ENREF_25)), but fires where the stand-replacing effects are concentrated in fewer large patches are much more susceptible to dispersal limitation of regenerating conifers compared to fires with similar percent high severity but more smaller patches ([Crotteau et al. 2013](#_ENREF_9), [Kemp et al. 2016](#_ENREF_17), [Welch et al. 2016](#_ENREF_41)). For instance, the 2013 Rim Fire in California’s Sierra Nevada had a relatively modest proportion of burned area mapped as high severity (~35%) but some of the largest contiguous patches of stand-replacing fire found anywhere in the modern record ([Lydersen et al. 2014](#_ENREF_19)). Thus, there is a need to update previous research on trends in the modern burn severity record by accounting explicitly for size and shape of stand-replacing patches ([Collins et al. 2017](#_ENREF_7)).

Our objective was to document trends in stand-replacing patch configuration in California’s mixed-conifer forest ecoregion over the past 33 years, using a novel metric developed to describe how much stand-replacing patch area remains with increasing distance inward from patch edges ([Collins et al. 2017](#_ENREF_7)). The stand-replacing decay coefficient (SDC) is related to fire size, high-severity area, and proportion high-severity, as well as conventional landscape metrics such as patch edge:area ratio ([Collins et al. 2017](#_ENREF_7)). However, this metric is more biologically relevant than the above metrics because it explicitly accounts for distance to seed source within stand-replacing patches, and as a single metric it distinguishes among fires that may be similar in terms of fire size or proportion high-severity but differ strongly in aggregate distance to seed source, without needing to specify a specific (and arbitrary) dispersal limitation distance ([Collins et al. 2017](#_ENREF_7)). Thus SDC can more directly identify fires that are vulnerable to long-term conifer forest loss and potential type-conversion. We present an updated analysis of the work by Miller and colleagues ([Miller and Safford 2012](#_ENREF_24), [Miller et al. 2012b](#_ENREF_26)) that includes fires through 2015, when California was in a historic multi-year drought, to investigate 1) whether fires with different managing agencies and management objectives differed in SDC independently of fire size and proportion high-severity, 2) how average SDC for these fires changed over time, and 3) the role of weather conditions in SDC. These results illustrate how a process-based quantification of fire effects can be used to describe changing fire regimes.

**Methods**

For our analysis we selected all wildfires in California that burned between 1984 and 2015 where the following criteria were met: 1) at least 80 ha in size, 2) predominantly (>50%) in yellow pine or mixed-conifer forest according to the CALVEG classification scheme ([Keeler-Wolf 2007](#_ENREF_16)), 3) occurring in northwestern California, the southern Cascades, or the Sierra Nevada, 4) predominantly (>50%) on land managed by either the US Forest Service or the US Park Service, and 5) having a mapped burn-severity classification layer available. These criteria led us to a sample size of 477 fires. For each fire we defined the location of stand-replacing fire the set of polygons mapped as >90% basal area mortality using the thresholds in Relative differenced Normalized Burn Ratio (RdNBR) from pre- and post-fire LANDSAT imagery described in [Miller et al. (2009a)](#_ENREF_22) and available at (https://www.fs.usda.gov/detail/r5/landmanagement/gis/?cid=stelprd3804878).

We calculated the stand-replacing decay coefficient (SDC) for each fire following the methods of Collins et al. (2017). SDC is defined by the following equation:

where *P* is the proportion of the original stand-replacing area in the fire that exceeds a given buffer distance inward from the patch edge (*D*), and *SDC* is a free parameter fit by nonlinear least squares estimation that simultaneously describes the size and complexity of stand-replacing area (Collins et al. 2017). We reasoned that not all edges are biologically equivalent, as outer edges of stand-replacing patches would be more likely to contribute conifer seed into the patch than edges of very small internal “holes” within stand-replacing patches that were mapped as ≤ 90% basal area mortality but most often were mapped as having > 75% basal area mortality. Therefore we filled in any “holes” of 9 contiguous 30 m pixels (0.81 ha) or smaller, and considered these part of the stand-replacing patch when calculating SDC.

For each fire we approximated the weather at the time of the fire using the GridMet database ([Abatzoglou 2013](#_ENREF_1)). We identified the start and end dates for each of our 477 fires; in rare cases where the end date was not known (N=35), we set the end date to seven days after the start date. We excluded cases where the start date was not known (N=4). We then calculated the centroid latitude and longitude coordinate of the high-severity area within a given fire, and downloaded the daily weather estimates from GridMet for the grid cell (4 km) overlapping that centroid during the burn period. Daily estimates were obtained for daily maximum temperature, minimum temperature, maximum relative humidity, and burn index (need to cite and explain what this is). For each fire we then identified the most extreme fire weather conditions for these four variables during the burn period (maximum high temperature (TMX), maximum low temperature (TMN), minimum high relative humidity (RH), and maximum daily burn index (BI), and incorporated these variables into our database of fires.

To evaluate the influence of weather and land management history/fire management (referred to hereafter as management) on variation in SDC, we compared a set of candidate models predicting SDC based on all possible combinations of seven variables, using automated model selection implemented in the R package *glmulti* ([Calcagno and de Mazancourt 2010](#_ENREF_5)). The variables examined were: fire year (1984-2015), fire management class (“class”; suppression or wildland fire use), management agency (CalFIRE, USFS, NPS), and the four weather variables (TMX, TMN, RH, BI). We selected the top 5 candidate models on the basis of AIC comparisons, and compared the parameter effect sizes across these models. With parameter effects consistent across the top five candidate models (Table 1), we selected a simple model (model #2) for a regression tree analysis using recursive partitioning, implemented in the *rpart* package in R ([Therneau et al. 2010](#_ENREF_40)).

**Results**

The best model to explain variation in SDC always included management class, management agency, and maximum high temperature during the burn window, while it never included the minimum high humidity (Table 1). Effects of these predictors were consistent: SDC decreased from NPS to USF to CDF-managed fires, decreased from WFU fires to suppression fires, and decreased with increasing maximum temperatures. Fire year, maximum burn index, and maximum low temperature were marginal additional predictors, with fire year always having a negative effect on SDC (Table 1). See Table 2 for a summary of weather and fire size statistics across different agencies and management classes.

The regression tree analysis indicated that the fire management class was a first-order control on SDC values, with higher SDC values – associated with smaller and/or more complex patches – for WFU fires (Figure 1). Non-WFU fires that were managed for suppression generally had lower SDC values that are associated with larger and/or simpler patches. Among non-WFU fires where the maximum high temperature during the burn window was less than 24 C, fires managed by the US Forest Service (N=26) had lower SDC values than fires managed by NPS (N=6) or CDF (N=3), which had the highest SDC values of any group of fires (-3.8, roughly equivalent to 1.1 ha circular patches; Figure 1, S1). Among non-WFU fires where the maximum high temperature during the burn window exceeded 24 C, the year of the fire was important, with fires occurring since 2010 having the lowest SDC values of any group of fires (-5.1, equivalent to roughly 12.5 ha circular patches; Figure 1, S1). Among non-WFU fires since 2010 where the maximum high temperature was greater than 24 C, fires with very high maximum high temperatures (>39 C) surprisingly had higher SDC values (Figure 1), while fires with maximum high temperatures between 24 and 39 C had lower SDC values if they were managed by CDF or USF, while if they were managed by the NPS their SDC values depended on temperature, with higher temperatures again leading to lower SDC values (Figure 1).

SDC is related to fire size and percent high-severity, because larger fires with more area burning at high-severity will inherently have more area located farther from high-severity patch edges (Collins et al. 2017). However, SDC provides additional information to distinguish fires from each other within a given range of fire size or percent severity. For instance, the reduction in SDC in fires managed by NPS or in fires managed as WFU fires are not just due to these fires being smaller in size or having lower percent high-severity (although these effects do exist). Rather, within a given fire size or percent high-severity range, agency and class still influence SDC (Figure 2). In a model of SDC conditional on class and either percent high-severity or fire size, class has a significant marginal effect on SDC after accounting for percent severity (t = 5.35, P < 0.001) and size (t = 7.92, P < 0.001). In a model of SDC conditional on agency and either percent high-severity or fire size, agency also has a significant marginal effect on SDC after accounting for percent severity, with NPS distinguishable from both USF and CDF but the latter two indistinguishable from each other.

While fire management class and agency are clearly related to SDC values, the relationship between fire year, weather during the fire, and SDC is more complex. SDC decreased over time (Figure 3), at a rate that was marginally significant for both the individual year averages (R2 = 0.11, t = 1.97, P = 0.058) and the five-year moving averages (R2 = 0.14, t = 2.08, P = 0.047). Interestingly, the trend in percent high severity over time was positive (consistent with the inverse relationship between SDC and percent high-severity), but not significant for individual year averages (R2 = 0.06, t = 1.43, P = 0.16) or five-year moving averages (R2 = 0.09, t = 1.62, P=0.12). The maximum high temperature, averaged across all fires within a given year, increased over time from 1984-2015 (Figure 3), a trend that was significant for the five-year moving average (R2 = 0.29, t = 3.29, P = .003) but not for individual year averages (R2 = 0.010, t = 1.83, P = 0.077). Similarly, the maximum average daily burn index increased over time (Figure 3), significantly both for individual year averages (R2 = 0.32, t = 3.80, P = 0.001) and for the five-year moving average (R2 = 0.69, t = 7.60, P < 0.001). However, while four of the six lowest average SDC values in the 31-year time period occurred between 2011 and 2015, only one of the six highest average burn index values occurred in this same period (Figure 3).

[I wonder if it is worth showing cumulative distributions of SDCs by agency (1984 to present) and then in the discussion try to tie that to “forest loss” over the last 30 years. We could pick a few thresholds for SDC that would approximate “loss” or near-term conversion to non-forest.]

**Discussion**

The stand-replacing decay coefficient (SDC) tended towards smaller values (e.g. larger and less complex patches) over time, in fires managed for suppression, and on landscapes with a longer history of fire suppression (e.g. USFS vs NPS), after accounting for the influence of weather. These broad trends are consistent with previous work documenting increases in the percentage of stand-replacing effects within a fire over time, and on Forest Service land rather than Park Service land ([Miller et al. 2009b](#_ENREF_25), [Miller et al. 2012a](#_ENREF_21), [Miller and Safford 2012](#_ENREF_24), [Miller et al. 2012b](#_ENREF_26)). However, in corroborating this previous work our results provide important additional information, because for the first time we are describing changes in the spatial patterns of stand-replacing fire that directly reflect changes in post-fire regeneration potential (e.g. distance to seed source).

The advantage of SDC over metrics such as percent high-severity is that fires with comparable percent high-severity can have dramatically different SDC values (Figure 2). This can be visualized in Figure 4, which presents a set of comparison fires with similar percent high-severity and similar fire area, but different spatial patterns and SDC values. SDC is useful as a single metric, comparable across a large number of fires, that simultaneously accounts for covariation in percent high-severity, area burned at high-severity, edge:area ratio of high-severity patches, and other metrics that are correlated with, but do not directly measure, the potential for dispersal limitation ([Collins et al. 2017](#_ENREF_7)). It is this dispersal limitation and lags in forest regeneration, rather than percentages of an area burning at high-severity *per se*, that contribute to potential forest loss and alternative stable states in the context climate change, increased fire frequency, and the potential for stand-replacing patches to re-burn at high-severity in the short term ([Millar and Stephenson 2015](#_ENREF_20), [Coppoletta et al. 2016](#_ENREF_8), [Harvey et al. 2016a](#_ENREF_11), [Johnstone et al. 2016](#_ENREF_14))

It is clear that both weather and fuels can strongly influence fire severity and area burned ([Safford et al. 2012](#_ENREF_33), [Collins 2014](#_ENREF_6), [Lydersen et al. 2014](#_ENREF_19), [Parks et al. 2015](#_ENREF_29)), and our results corroborate this for spatial patterns of stand-replacing fire as well. We see more fire effects within the range of historical variability for California mixed-conifer forests – smaller, more irregular patches of stand-replacing fire ([Safford and Stevens 2017](#_ENREF_32)) – under more moderate weather conditions, with maximum daily high temperature during the burn period emerging as an important factor (Figure 1). Although management class emerged as the first-order control over SDC (Figure 1), this also reflects the influence of weather to some degree, as “wildland fire use” fires tend to burn under cooler maximum high temperatures than fires managed for suppression (Table 2). Similarly, fires in the National Parks tend to have cooler maximum high temperatures than fires on National Forest land, even when suppression is the management objective (Table 2).

While we do not account for fuels directly in our analysis, several lines of evidence suggest that increased fuel loads are associated with lower SDC values. The trend towards lower SDC values over time may signal of the effect of fire suppression and associated fuel accumulation. The years from 2011-2015 had four of the six lowest mean SDC values of any year since 1984, and while maximum temperature and burn index increased over this time period, only two of those years (2012 and 2015) were among the six highest maximum temperature years, and only one (2012) was among the six highest burn index years (Figure 3). Our regression tree analysis identifies 2010 as a threshold year, with fires occurring on or after that year having the lowest mean SDC value of any cluster in the tree, after controlling for the effect of temperature (Figure 2). Furthermore, the lower values of SDC for fires managed by the US Forest Service than the Park Service after controlling for weather (Figure 2) may indicate a longer history of fire suppression on Forest Service lands ([Miller et al. 2012a](#_ENREF_21)), which have a broader array of constraints when considering how to manage ignitions (cite).

Topography is also an important control over fire effects ([Taylor and Skinner 2003](#_ENREF_39), [Lydersen et al. 2014](#_ENREF_19), [Harris and Taylor 2015](#_ENREF_10)). In areas with high topographic complexity, patterns of stand-replacing fire may be less responsive to variation in fuels or weather ([Miller et al. 2012b](#_ENREF_26)). We found a seemingly counterintuitive result in our regression tree analysis where fires with a maximum high temperature greater than or equal to 39°C had smaller SDC values (N=18, Figure 2), however every one of these fires occurred in the northwestern part of California centered around the Klamath Mountains, with a majority (N=10) occurring in 1987, which was a particularly warm year (Figure 3) with widespread fire activity in this region. The Klamath Mountains have high topographic complexity, which likely induces some limitation to the spatial extent of stand-replacing fire (e.g. on steep North-facing slopes), even under warm conditions ([Taylor and Skinner 2003](#_ENREF_39), [Miller et al. 2012b](#_ENREF_26)).

While it is difficult to ascribe strict causality to the observed trends in SDC, multiple lines of evidence suggest that SDC is responding as expected to variation in weather, fuels and topography, across both space and time. It is highly likely that these three factors act in concert to produce larger, more regular patches of stand-replacing fire under certain conditions, particularly under extreme fire weather conditions on landscapes that have an accumulation of fuel and relatively simple topography. Further, it is highly likely that conditions are changing over time towards increasingly extreme fire weather and increasingly fuel-loaded landscapes ([Millar and Stephenson 2015](#_ENREF_20), [Safford and Stevens 2017](#_ENREF_32)), which suggest that the occurrence of so-called “mega-fires”, where fire behavior and effects exceed the range of variability previously observed, will continue to increase over time ([Stephens et al. 2014](#_ENREF_35)). Low SDC appears to be a good indicator “mega-fires”, and their incidence appears to be on the rise. Over the 32 years from 1984-2015, 20 fires have had an SDC lower than 0.0026. Of these 20 fires, half (10) have occurred in the 9 years since 2007, including some well-known recent fires widely considered to be “mega-fires”, including the 2007 Moonlight Fire (citation?), the 2013 Rim Fire ([Lydersen et al. 2014](#_ENREF_19)), and the 2015 King Fire ([Jones et al. 2016](#_ENREF_15)), which has the lowest SDC of any of the 477 fires studied (SDC = 0.0013; ln(SDC) = -6.64).

Fires with more desirable SDC values (e.g. SDC > 0.0067; ln(SDC) = -5; Figure 3, 4) suggest a way forward for fire management that incorporates some of the benefits of stand-replacing fire while not compromising long-term forest resilience. Managed wildfires that burn under moderate fire weather conditions or landscapes with a past history of fire use or other fuel management are much more likely to have smaller SDC values (Figure 2). This is consistent with a large and developing body of literature suggesting that there are opportunities for increased use of fire, in concert with mechanical fuels reduction in some instances, during periods of time where rapid fire spread is not likely ([Stephens et al. 2013](#_ENREF_34), [Millar and Stephenson 2015](#_ENREF_20), [North et al. 2015](#_ENREF_28), [Stephens et al. 2016](#_ENREF_36)). There are many barriers to the increased use of fire, but current trends in the spatial patterns of stand-replacing mean that the alternative could be increasingly large stand-replacing patches subject to forest loss for extended periods of time.

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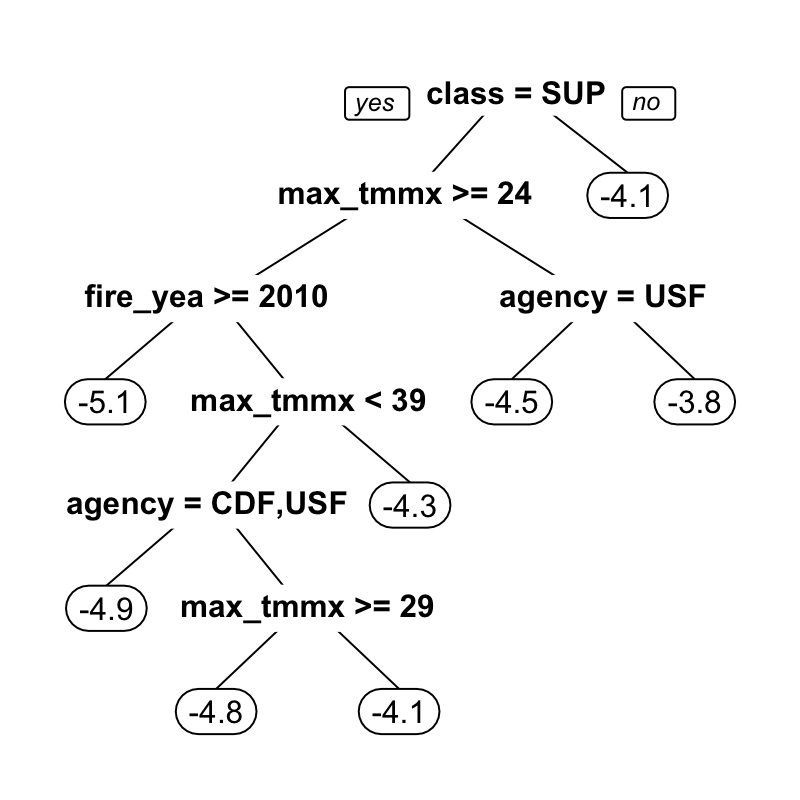
**Table 1**: Five best candidate models of SDC, based on AIC comparison.

|  | **Model #** | | | | |
| --- | --- | --- | --- | --- | --- |
| **Model AIC  /coefficients** | **1** | **2** | **3** | **4** | **5** |
| AIC | 890.12 | 890.73 | 890.86 | 890.87 | 891.17 |
| (Intercept) | 6.645 | 4.993 | -4.36 | 4.741 | -4.502 |
| agencyUSF | 0.386 | 0.387 | 0.422 | 0.412 | 0.42 |
| agencyNPS | 0.483 | 0.512 | 0.481 | 0.475 | 0.508 |
| classWFU | 0.193 | 0.211 | 0.176 | 0.185 | 0.195 |
| max\_tmmx | -0.006 | -0.005 | -0.02 | -0.005 | -0.009 |
| fire\_year | -0.022 | -0.01 |  | -0.022 |  |
| max\_bi |  |  | 0.018 | 0.02 | -0.003 |
| max\_tmmn | 0.019 |  | -0.004 | -0.003 |  |
| min\_rmax |  |  |  |  |  |

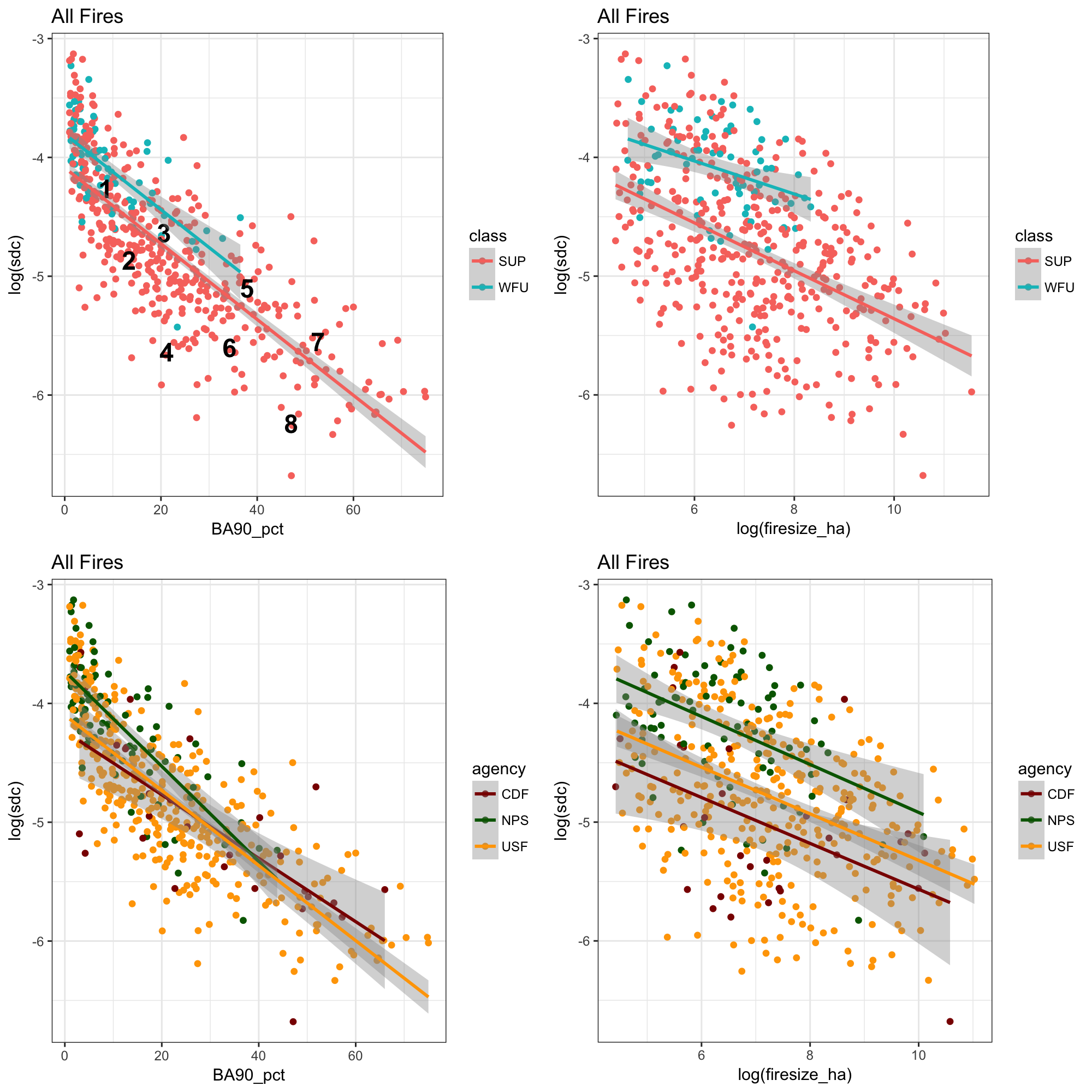
**Table 2**: Summary of fire statistics across agency and management class. Fires with agency = NA were other agencies not among the three principal fire management agencies and with too few fires to draw meaningful conclusions (e.g. Bureau of Indian Affairs)

| **agency** | **class** | **N** | **min size  (ha)** | **median  size (ha)** | **max size  (ha)** | **median  fire year** | **mean maximum  high temperature** | **mean maximum burn index** | **mean maximum  low temperature** | **mean minimum  high humidity** |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| CDF | SUP | 31 | 83 | 828.0 | 39265 | 2000 | 32.7 | 70.9 | 14.5 | 40.0 |
| NPS | SUP | 33 | 84 | 414.0 | 24123 | 1996 | 27.8 | 67.7 | 12.7 | 36.4 |
| NPS | WFU | 54 | 106 | 642.5 | 4143 | 1996 | 25.9 | 74.3 | 11.7 | 29.1 |
| USF | SUP | 335 | 85 | 1385.0 | 61516 | 2003 | 32.5 | 69.1 | 15.5 | 37.5 |
| USF | WFU | 17 | 140 | 928.0 | 2420 | 2003 | 24.9 | 79.3 | 10.4 | 28.2 |
| *NA* | SUP | 7 | 102 | 742.0 | 104038 | 2008 | 27.8 | 71.9 | 13.7 | 36.2 |

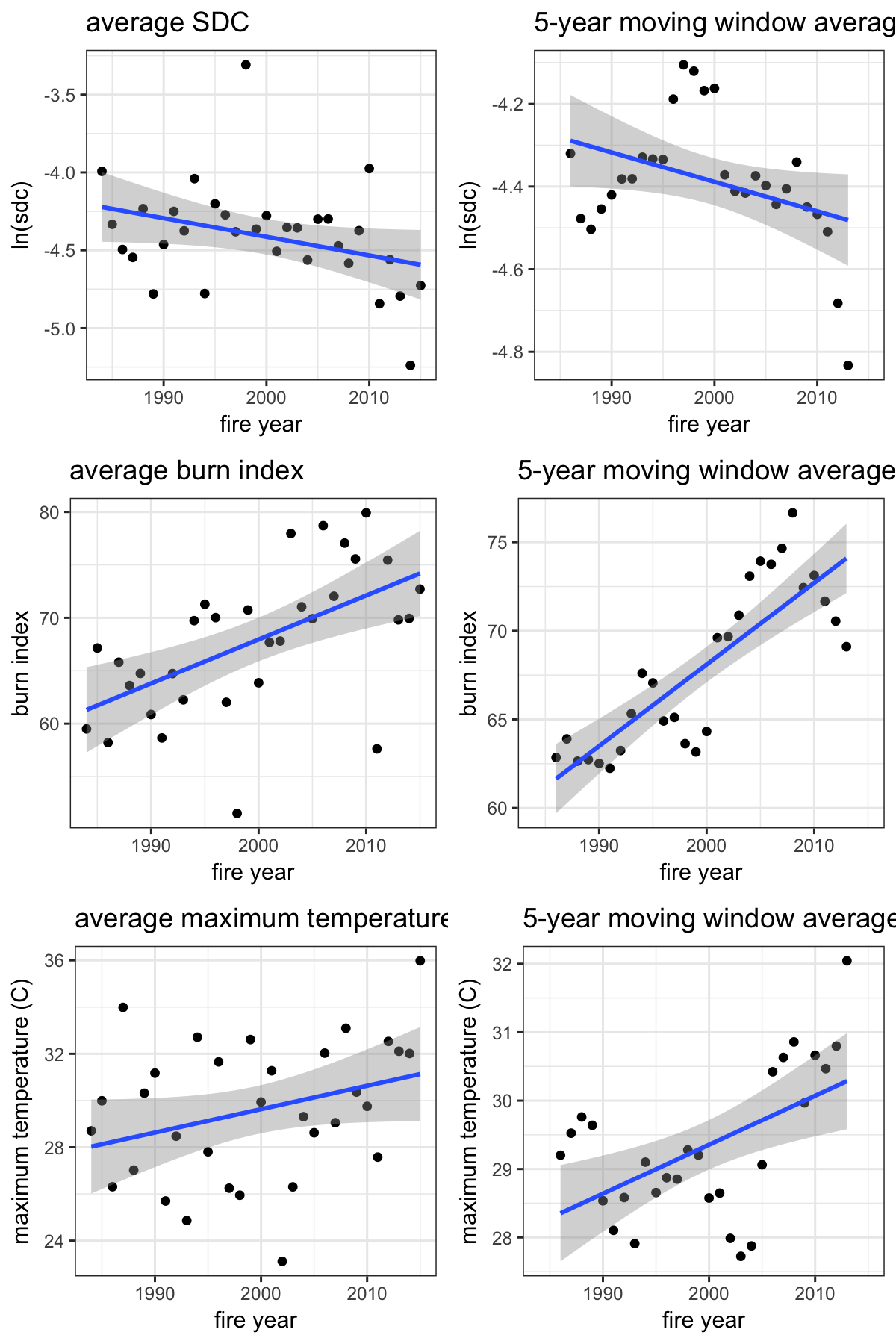
**Figure 1**: Regression tree based off model 2 (Table 1). Values in ovals are ln-transformed SDC values. Variables are fire management class (Suppression *SUP* or Wildland fire-use *WFU*), max high temperature during the burn window (max\_tmmx), fire year (1984 through 2015), and fire management agency (National Park Service *NPS*, US Forest Service *USF*, CalFIRE *CDF*).



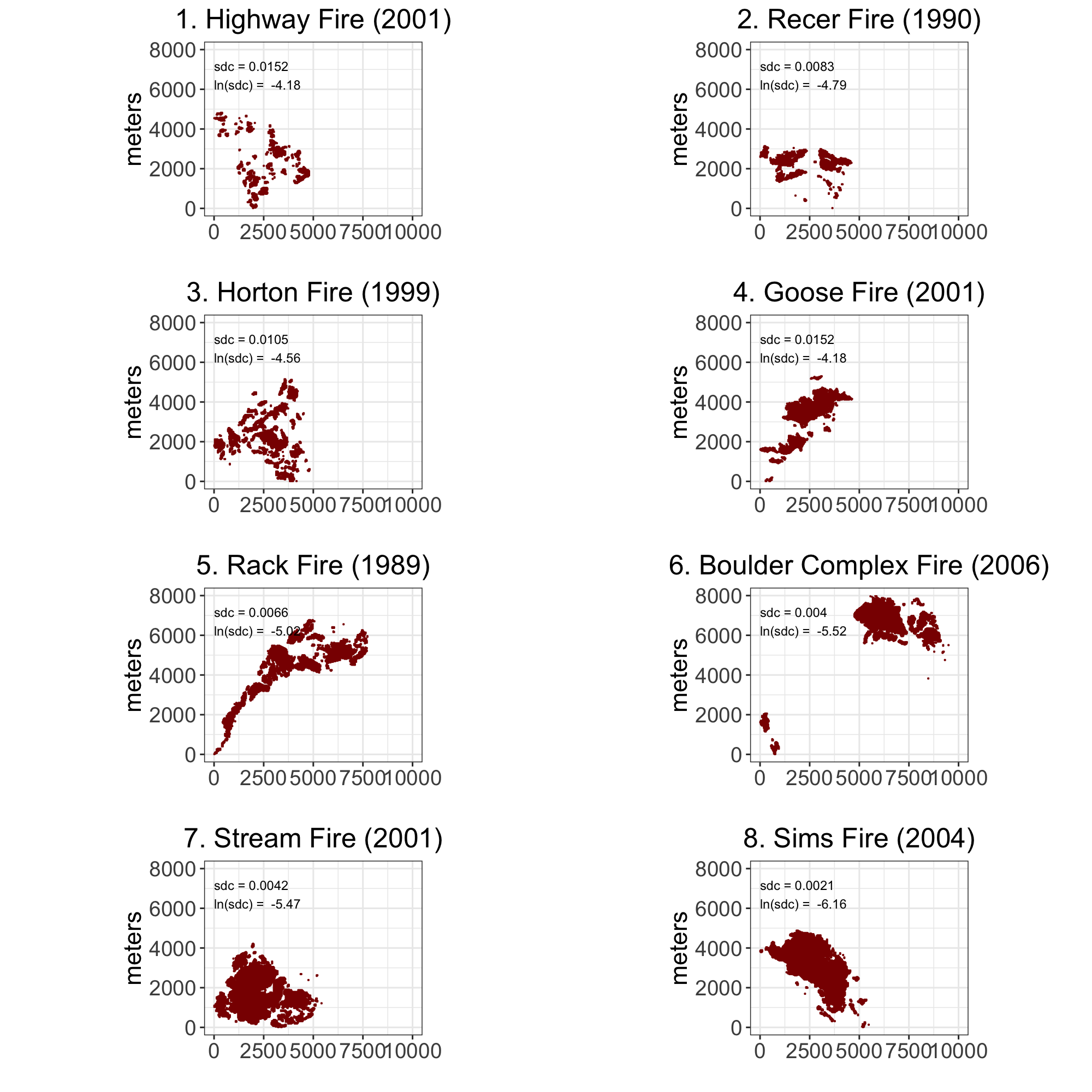
**Figure 2:**



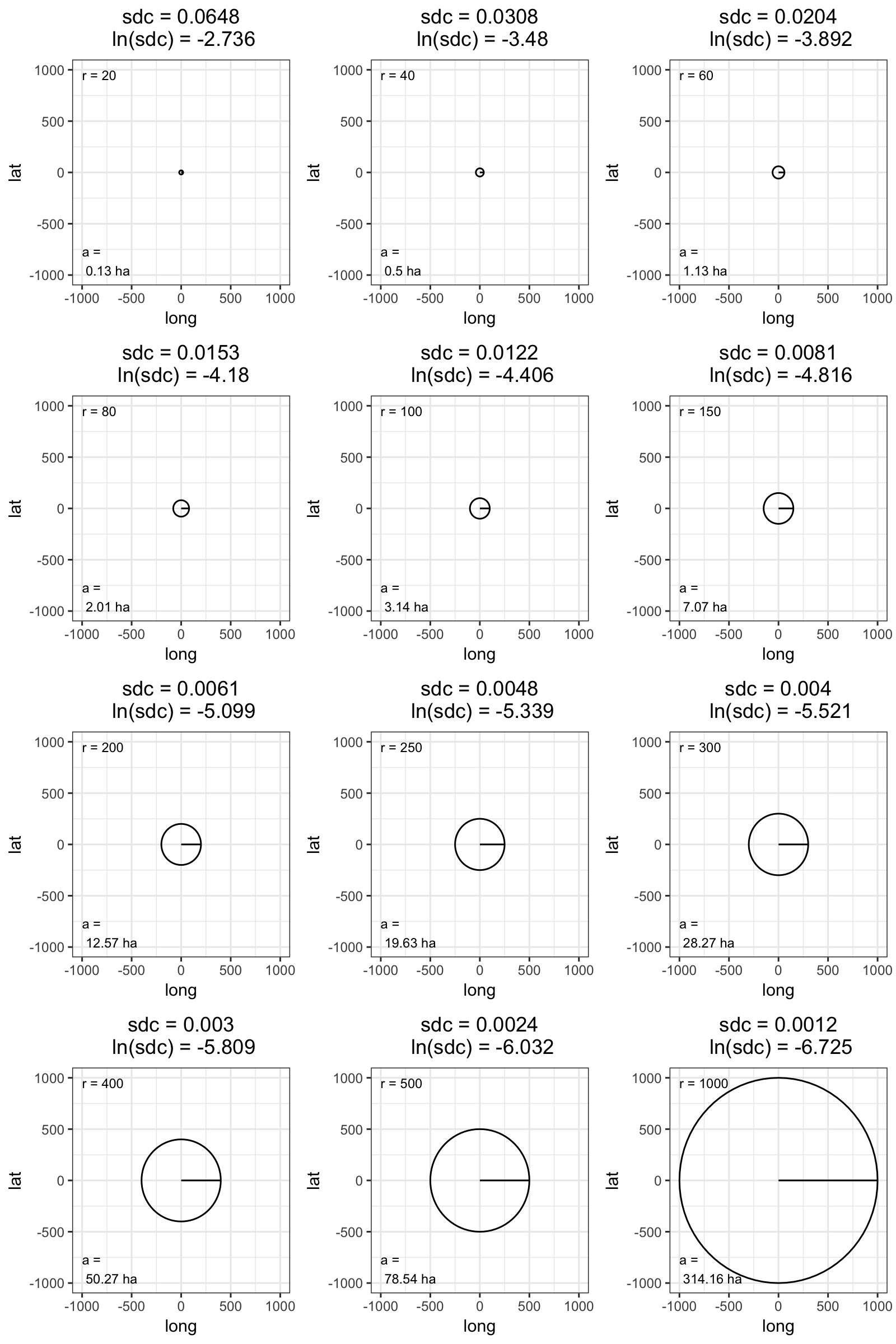
**Figure 3:** Trends in SDC, burn index and maximum high temperature over time.



**Figure 4**: Examples of SDC for a range of fires. Fires in the same row have similar areas and percent high-severity, corresponding to numbers 1-8 in Fig. 3. (Details will be filled in). SDC values are shown on figure. Fires in the right column have lower SDC values than comparably-sized fires in the left column.



**Figure S1**: Range of possible SDC values as a function of patch radius

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