



Season Spotter: Using Citizen Science to Validate and Scale Plant Phenology from Near-Surface Remote Sensing

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1 Article

2 Season Spotter: Using citizen science to validate and 3 scale plant phenology from near-surface remote 4 sensing

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14 **Abstract:** The impact of a rapidly changing climate on the biosphere is an urgent area of research
15 for mitigation policy and management. Plant phenology is a sensitive indicator of climate change
16 and regulates the seasonality of carbon, water, and energy fluxes between the land surface and the
17 climate system, making it an important tool for studying biosphere-atmosphere interactions. To
18 monitor plant phenology at regional and continental scales, automated near-surface cameras are
19 being increasingly used to supplement phenology data derived from satellite imagery and data
20 from ground-based human observers. We used imagery from a network of phenology cameras in
21 a citizen science project called Season Spotter to investigate whether information could be derived
22 from these images beyond standard, color-based vegetation indices. We found that engaging
23 citizen science volunteers resulted in useful science knowledge in three ways: first, volunteers
24 were able to detect some, but not all, reproductive phenology events, connecting landscape-level
25 measures with field-based measures. Second, volunteers successfully demarcated individual trees
26 in landscape imagery, facilitating scaling of vegetation indices from organism to ecosystem. And
27 third, volunteers' data were used to validate phenology transition dates calculated from
28 vegetation indices and to identify potential improvements to existing algorithms to enable better
29 biological interpretation. As a result, the use of citizen science in combination with near-surface
30 remote sensing of phenology can be used to link ground-based phenology observations to satellite
31 sensor data for scaling and validation. Well-designed citizen science projects targeting improved
32 data processing and validation of remote sensing imagery hold promise for providing the data
33 needed to address grand challenges in environmental science and Earth observation.

34 **Keywords:** citizen science; crowdsourcing; phenology; PhenoCam; landscape ecology; spatial
35 scaling, vegetation indices
36

37 1. Introduction

38 Plant phenology, the timing of life history events such as leaf-out, flowering,
39 seed-development, and senescence, is highly sensitive to weather, and is thus a key indicator of the
40 impacts of climate change on Earth's biota [1]. Warming spring temperatures over the past half
41 century have caused plant species across the temperate zone to leaf out earlier [2,3]. Likewise,
42 delayed autumn chilling has widely delayed leaf senescence [2–4], though the timing of spring
43 phenology modifies this effect [5].

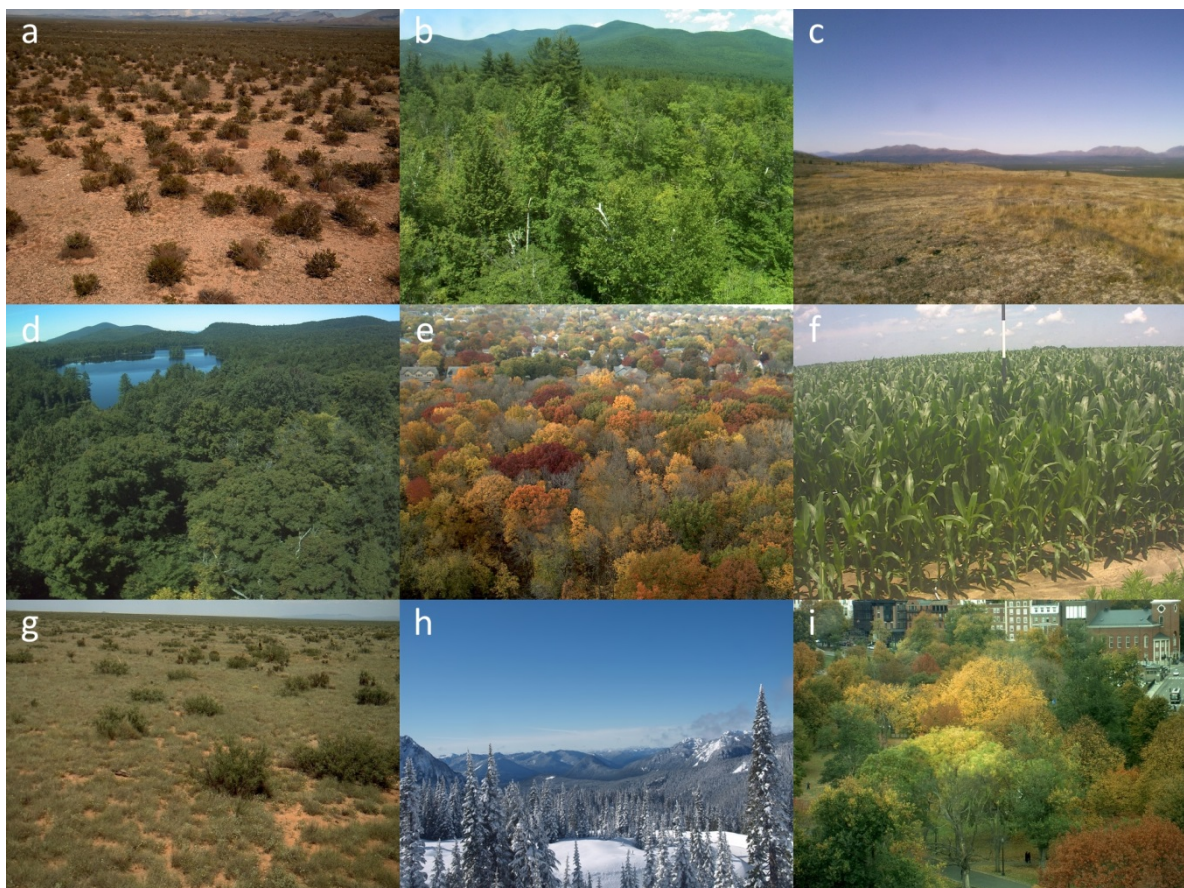
44 Currently, plant phenology is studied on the ground at the scale of the individual [6] or at
45 broader scales using vegetation indices derived from satellite imagery [7] or near-surface automatic
46 digital cameras [8]. Vegetation indices are typically used to track continuous photosynthetic activity
47 of vegetation, whereas ground-based field measurements are often needed to determine discrete
48 phenological events, including reproductive phenology [9,10]. Estimates of the effects of climate

49 change on plant phenology using data at different scales vary [11], and so there is need to integrate
 50 phenology measures at different scales.

51 Midscale phenology cameras can provide a link between these different scales of observation
 52 by extracting information beyond vegetation indices from the camera imagery [10]. Because the
 53 spatial scale of these near-surface images is much finer than that of satellite imagery, details such as
 54 the presence or color of flowers, the foliage and color of individual tree canopies, and the
 55 occurrence of precipitation events (including rain, fog and snow) can be seen directly in camera
 56 images. Documenting the phenological state of individuals is typical of on-the-ground phenology
 57 data gathering efforts. However, phenology cameras have rarely been used to produce information
 58 beyond landscape-scale leaf canopy state (but see [12]).

59 Extracting these sorts of details—rather than general patterns of “vegetation greenness”—from
 60 phenology camera images requires sophisticated image analysis that has not yet been automated,
 61 and the sheer volume of imagery makes manual image analysis prohibitive for networks of more
 62 than a handful of cameras. Citizen science provides a solution for analyzing large image data sets to
 63 extract information that cannot yet be easily extracted by computational means [13]. Citizen
 64 scientist volunteers have successfully classified galaxy types from astronomical telescope images
 65 [14], found interstellar dust particles in electron microscope images [15], and identified animal
 66 species in camera trap images [16].

67 We created the citizen science project Season Spotter (seasonspotter.org) to analyze images
 68 from phenology cameras. Season Spotter uses images from the PhenoCam network, the largest
 69 near-surface phenology camera network in the world. It consists of 300 elevated cameras located
 70 primarily in North America and spans a wide range of ecosystem types (Figure 1) [8,17].
 71 Automated processing of PhenoCam images produces a vegetation index (G_{CC} , “green chromatic
 72 coordinate”) that indicates the relative amount of green in a predefined region of interest in each
 73 image [18]. This G_{CC} index can then be used to infer the seasonal progression of leaf emergence and
 74 expansion, leaf color change, and senescence in the same manner as satellite-sensor-derived
 75 vegetation indices like NDVI and EVI [19]. G_{CC} has also been shown to mirror the dynamics of
 76 carbon dioxide fluxes as measured by co-located eddy covariance instrumentation [20].



78 **Figure 1.** Examples of landscape images as seen by phenology cameras in the PhenoCam network.
 79 Shown are sites (a) jerbajada; (b) bartlettir; (c) snipelake; (d) arbutuslake; (e) downerwoods; (f)
 80 uiefmaize; (g) ibp; (h) mountranier; (i) bostoncommon. See Table 1 for site descriptions.

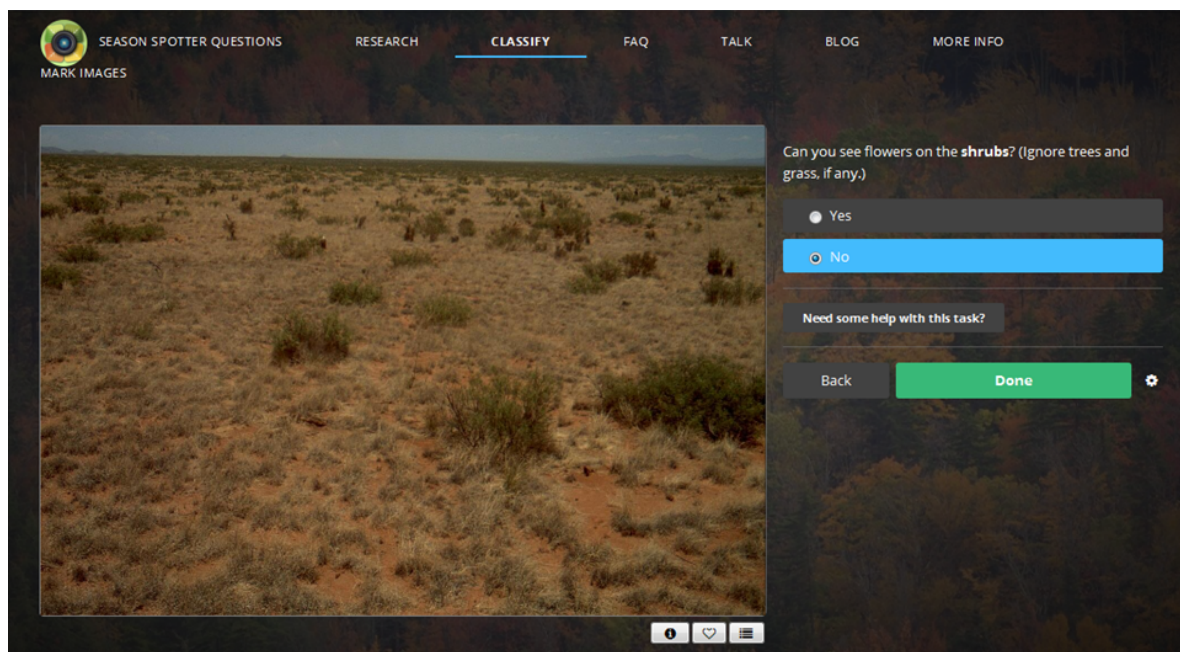
81 Season Spotter presented PhenoCam images from across a wide range of ecosystem types to
 82 volunteers using an online interface and asks the volunteers questions about these images. We
 83 algorithmically combined the volunteers' answers into usable classifications. Season Spotter asked
 84 volunteers to: identify the reproductive and vegetative states of deciduous trees, evergreen trees,
 85 shrubs, grasses, forbs, and crops; identify poor quality images and those containing snow; outline
 86 individual trees at forested sites; and make phenological comparisons between two images taken at
 87 the same site.

88 Our goals for leveraging human visual perception using Season Spotter were threefold. First,
 89 we wanted to discern whether reproductive phenology (e.g. flowers, fruits, seeds) could be detected
 90 from these images to provide a complementary data product connecting landscape-level phenology
 91 measurements with field-based measures. Second, we wanted to see if individual trees could be
 92 identified and their vegetation indices calculated to facilitate scaling from local to regional scales.
 93 And third, we wanted to use citizen scientists' assessments of spring and autumn start and end
 94 dates to evaluate dates calculated automatically from vegetation indices.

95 2. Materials and Methods

96 2.1. Citizen Science: Season Spotter

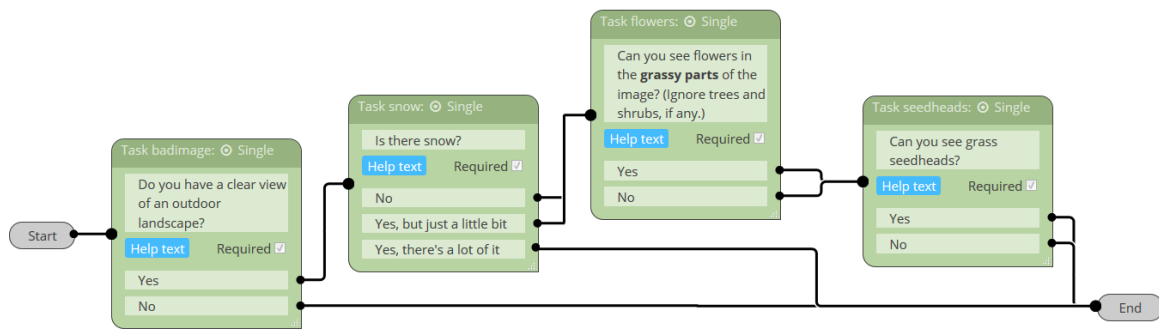
97 We created the online citizen science project Season Spotter (seasonspotter.org) using the
 98 Zooniverse Project Builder (www.zooniverse.org/lab) [21]. The Zooniverse is an online citizen
 99 science platform with 1.1 million current users, which hosts a variety of projects in need of
 100 volunteers to support data processing tasks. The Season Spotter site consists of a landing page,
 101 which allows volunteers to choose whether they want to answer multiple-choice questions about
 102 PhenoCam images or demarcate specified regions on the images (Figure 2).



103
 104 **Figure 2.** The Season Spotter user interface

105 After making the choice, each volunteer begins one of multiple randomly-assigned workflows
 106 (Figure 3, S1) that are tailored to the different ecosystem types. Each workflow consists of one or
 107 more tasks, including answering questions and drawing outlines. When a volunteer begins a
 108 workflow, a PhenoCam image is selected and presented to the volunteer along with its associated

109 tasks. For each task, there is a help button that, when clicked, provides the volunteer with detailed
 110 instructions for the task to be completed together with example images. When the volunteer has
 111 finished answering questions and/or demarcating regions, the project shows a summary of the
 112 volunteer’s responses and a button to load a new image.



113

114 **Figure 3.** Workflow for PhenoCam images containing grass. See Supplementary Materials Figure
 115 S1 for all other workflows.

116 The summary also provides a button for entering a dedicated chat forum where volunteers can
 117 ask questions, comment on images, and interact with the Season Spotter science team and each
 118 other. Additional outreach and engagement is regularly conducted via the Season Spotter blog
 119 (seasonspotter.wordpress.com), Facebook (www.facebook.com/seasonspotter), and Twitter
 120 (twitter.com/seasonspotter).

121 In May 2015 we tested the Season Spotter project with a group of 39 volunteer beta testers.
 122 Using their feedback on a follow-up questionnaire, we modified task questions and added
 123 additional instructions and information to increase data quality. On July 21, 2015, we officially
 124 launched Season Spotter. The majority of volunteer recruitment occurred through Zooniverse email
 125 newsletters, though we undertook multiple recruitment activities [22].

126 *2.2. PhenoCam Images*

127 We uploaded 51,782 images to the Season Spotter project, divided into three groups. Each
 128 image came from a PhenoCam site (

129 Table 1, Figure 4) and had its top or bottom cropped off to remove date information as well as
 130 to provide a more visually pleasing experience. The cropped images ranged from 640x452 pixels to
 131 3888x2592 pixels, and image viewing resolution varied among volunteers by device, operating
 132 system, and browser.



133

134 **Figure 4.** Map of PhenoCam sites used in Season Spotter. See

135

Table

1

for

site

descriptions.

