1	MOD-LSP v1.0: MODIS-Based Land Surface Parameters for the
2	Variable Infiltration Capacity Model over the Continental US,
3	Mexico, and Southern Canada
4	Dataset Description, Methods and User Guide
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#### 48 1. Introduction

This document describes the MOD-LSP MODIS-based land surface parameters for the 49 Variable Infiltration Capacity (VIC) model (Liang et al. 1994) compatible with release 5.0 and 50 51 later (Hamman et al. 2018) in image mode. The MOD-LSP spatial domain covers the continental 52 United States, Mexico, and southern Canada. This spatial domain and 0.625° (6 km) grid resolution are compatible with the gridded daily meteorological forcings of (Livneh et al. 2015), 53 54 which can be disaggregated to hourly time step via the MetSim tool (Bennett et al. 2018) using 55 domain files developed for PITRI precipitation disaggregation (Bohn et al. 2019). These domain 56 files also can accompany the MOD-LSP parameter files as inputs to VIC simulations.

These parameters have two main purposes: (1) to improve upon previous widely-used parameters over the region with updated, higher-resolution land cover maps and spatially explicit observations of surface properties; and (2) to expand from a single parameter set corresponding to one point in time to a series of parameter sets that account for temporal variability at seasonal to decadal scales.

The previous generation of VIC model parameters over the United States (Maurer et al. 2002; Mitchell et al. 2004; Livneh et al. 2013), Mexico (Zhu and Lettenmaier 2007), and North America (Livneh et al. 2015) share several limitations: (1) they were derived from a relatively coarse-resolution land cover map, based on AVHRR imagery acquired in the early 1990s (Hansen et al. 2000); (2) the annual cycle of monthly leaf area index (LAI) values was derived by spatially interpolating a sparse (< 100 points over North America) set of pixels from a single year of an AVHRR-based dataset (Myneni et al. 1997); (3) in all land cover classes except

explicit bare soil, the vegetation canopy was assumed to have 100% area coverage (no gaps); (4)
each land cover class was assigned a single spatially (and temporally, for most classes) invariant
value of albedo obtained from literature; and (5) urban areas were treated as bare soil.

72 In addition, the prior parameter sets did not account for land cover variability and change, including: (1) long-term land cover conversions in response to climate change, disturbance, and 73 74 land use; and (2) interannual variations in phenology (plant characteristics such as LAI and albedo) in response to climate fluctuations. Both types of change occur at varying rates on the 75 landscape (e.g., withering and eventual death of trees in response to drought; Breshears et al. 76 2005). The land cover map in prior VIC parameter sets represented land surface conditions circa 77 78 1992-3. Because no subsequent maps with similar methodology were made, the parameters were not suitable for studies of land cover change. In the subsequent two decades, many land cover 79 change datasets, consisting of land cover classifications using the same methodology and spaced 80 several years apart, have been developed at a variety of scales. In particular, the Multi-81 82 Resolution Land Characteristics Consortium (MLRC) National Land Cover Database (NLCD; 83 Homer et al. 2015) contains consistent classifications over the United States for years 1992, 84 2001, 2006, 2011, and 2016; and the Instituto Nacional de Estadística y Geografía (INEGI) Uso 85 del Suelo y Vegetación land cover product (INEGI; INEGI, 2014) contains consistent classifications over Mexico for years 1985, 1993, 2002, 2007, and 2011. Similarly, interannual 86 variability in phenology, as observed by the Moderate Resolution Imaging Spectroradiometer 87 (MODIS), has recently received attention in hydrologic modeling studies (Hogue et al. 2005; 88 89 Zhang et al. 2013; Tang et al. 2012; Ford and Quiring 2013; Parr et al. 2015; Tesemma et al. 90 2015; Bohn and Vivoni 2016; Liu et al. 2018).

MOD-LSP was developed to overcome the limitations of prior parameter sets and 91 facilitate studies of land cover change, over the continental United States, Mexico, and southern 92 93 Canada. To facilitate land cover change analysis, several versions of the MOD-LSP parameters have been generated using harmonized NLCD and INEGI classifications from years 1992/3, 94 2001/2, and 2011, as well as a land cover classification based on MODIS. In terms of phenology, 95 96 MOD-LSP introduces the following improvements: (1) the annual cycle of phenology has been derived from 17 years (2000-2016) of 8- and 16-day MODIS products at nearly all of the 500 m 97 pixels in the domain, yielding spatially explicit and statistically representative estimates of 98 phenology for each land cover class in each grid cell; (2) taking advantage of recent VIC model 99 development (Bohn and Vivoni 2016), canopy fraction (*f<sub>canopy</sub>*) and albedo have also been 100 estimated from MODIS products in similar fashion to LAI; and (3) values have been provided 101 for urban areas. Alternate versions of the annual cycles of phenology derived from a single year 102 103 corresponding to the land cover classification have also been generated. Finally, to account for 104 interannual variability in phenology, monthly time series of phenology spanning the period 2000-2016 have been generated as optional vegetative forcings for the VIC model. 105

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#### 107 2. Methods

108 2.1. Overview

Processing of the land surface properties centers around spatial averaging of the MODIS phenology observations over the classes of the chosen land cover classification within each output grid cell, yielding a separate spatial average for each class (Fig. 1). Before aggregation could take place, the NLCD and INEGI land cover classifications needed to be harmonized. After aggregation, data gaps were filled via temporal and spatial interpolation. Finally, to be

usable in VIC simulations, other non-MODIS land surface properties were copied from prior
parameter sets, mapped from old to new land cover classes where necessary.

116 2.2. Study Domain

The study domain spans the region bounded by 14.5° and 53° N latitude and 125° and 67° W longitude (Fig. 2). This domain (also used in Livneh et al., 2015) is an extension of the CONUS domain, over which several gridded meteorology and hydrology datasets (Maurer et al. 2002; Mitchell et al. 2004; Xia et al. 2012; Livneh et al. 2013) have been constructed, to include the nation of Mexico; hence we will refer to the domain as CONUS\_MX hereafter. In those cases for which our analysis omits southern Canada, we will refer to that subset of the domain as USMX.

124 2.3. VIC Parameters

125 2.3.1. General Information

Several spatial datasets were used in creating and evaluating the parameter datasets under 126 127 consideration herein. Details of satellite sensors, spatial resolution, and acquisition dates for these datasets can be found in Table 1. The VIC parameter datasets discussed herein (described 128 in Table 2) include "L2015", the parameter set used in Livneh et al. (2015), covering the 129 CONUS\_MX domain (obtained from http://ciresgroups.colorado.edu/livneh/data/daily-130 observational-hydrometeorology-data-set-north-american-extent); and "MOD-LSP", the suite of 131 new parameter sets with phenology based on MODIS observations, covering either CONUS\_MX 132 or USMX domains, depending on the underlying land cover classifications. All VIC parameter 133 datasets were created at 0.0625° (6 km) spatial resolution. 134

#### 135 2.3.2. Land Cover Classifications

The land cover classification underlying the L2015 dataset was the AVHRR-based
University of Maryland (UMD) land cover product (Hansen et al. 2000), modified for the North
American Land Data Assimilation project (NLDAS) (Mitchell et al. 2004) to exclude the Open
Water, Urban, and Perennial Ice/Snow classes (UMD-NLDAS hereafter). In each grid cell, the
Urban and Perennial Ice/Snow classes were replaced by bare soil, and the Open Water class was
replaced by all other classes present in the grid cell in proportion to their area fractions.

The MOD-LSP parameter sets are named based on the underlying land cover 142 143 classification, according to the convention: lc source.lc year, where lc source indicates the land cover source and methodology and *lc\_year* indicates the year or set of years to which the land 144 cover classification pertains. Values of *lc\_source* include: MOD\_IGBP, based on the IGBP 145 146 classification of the MOD12Q1.005 land cover product (Friedl et al. 2010), covering the CONUS MX domain; and NLCD INEGI, which covers the USMX domain and is based on the 147 148 National Land Cover Database (NLCD; Homer et al., 2015) over the United States and the Uso 149 del Suelo y Vegetación land cover product of the Instituto Nacional de Estadística y Geografía (INEGI; INEGI, 2014) over Mexico. To combine the NLCD and INEGI products into a single 150 product, the more numerous INEGI land cover classes were mapped to the 16 NLCD 2011 151 classes, as shown in Table 3. Details of the procedure can be found in the NLCD INEGI GitHub 152 archive (https://github.com/tbohn/NLCD\_INEGI/releases/tag/v1.5) (Bohn 2019a). 153 154 Because the MOD12Q1.005 land cover product provided 13 separate annual maps over the period 2001-2013, during which many pixels oscillated back and forth among several classes, 155 each pixel in the MOD\_IGBP parameter set was assigned the most frequent land cover class 156

157 encountered over the 13-year period, yielding a single map that represented the entire period.

This has been denoted by setting *lc\_year* to "mode". The NLCD\_INEGI datasets used the 158 National Land Cover Database (NLCD) maps from 1992, 2001, and 2011 (Homer et al. 2007; 159 160 Fry et al. 2009; Homer et al. 2015) over the US and the Uso del Suelo y Vegetación maps from 1993, 2002, and 2011 (INEGI 2014) over Mexico. It should be noted that although both the 161 162 NLCD and INEGI products were derived from Landsat imagery, they were created with different 163 methodologies: NLCD was classified automatically on a pixel-by-pixel basis, while the INEGI product was delineated manually into polygons of various classes. Furthermore, the NLCD 1992 164 product used a different method from that used in subsequent years. The MRLC issued a 1992-165 2001 land cover change retrofit product (Fry et al. 2009) to facilitate comparison between those 166 years. However, the retrofit product only described changes in broad Anderson Level I land 167 cover categories (e.g., "forest" or "shrub/grassland"). We created a version of the NLCD 1992 168 map that was both consistent with the retrofit and contained information about all classes as 169 follows: where the 1992-2001 retrofit product indicated that land cover change had occurred, 170 171 classes at Anderson Level II (e.g., "deciduous forest" or "shrubland") were taken from the NLCD 1992 classification (with 1992 class codes mapped to the 2001 legend); otherwise classes 172 were taken from the NLCD 2001 classification (Homer et al. 2007). For the NLCD INEGI 173 174 parameter sets, *lc\_year* was set to the year of the NLCD product used (1992, 2001, or 2011), with the INEGI products from 1993 and 2002 used in *lc\_years* 1992 and 2001. The *lc\_year* code 175 was prepended with "s" in some cases to indicate that only "stable" pixels (those that did not 176 177 change land cover class over the period 2001-2011) were used to compute spatial average values of phenology variables (see below). Details of the procedure can be found in the NLCD\_INEGI 178 179 GitHub archive (https://github.com/tbohn/NLCD\_INEGI/releases/tag/v1.5) (Bohn 2019a).

180 2.3.3. Time-Varying Surface Properties

The VIC model requires specification of a repeating annual cycle of 12 monthly values of the following time-varying surface properties (phenology) for each land cover class of each grid cell: Leaf Area Index (LAI), albedo, roughness length, and displacement height. An additional annual cycle of canopy fraction ( $f_{canopy}$ ) is optional, and was included in the MOD-LSP parameter sets. For roughness length and displacement height, all VIC parameter datasets used a spatiallyinvariant repeating annual cycle of monthly values for each land cover class, taken from literature (Nijssen et al. 2001).

188 The L2015 parameter set contained a unique annual cycle of monthly LAI values for each land cover class in each grid cell, based on the AVHRR-based product of Buermann et al. (2002) 189 (B2002 hereafter). This annual cycle was derived from only a single year of values from 1992-190 191 1993, and from a sparse (< 50) set of pixels from each land cover class, which were extended over the entire CONUS MX domain by spatial interpolation. For albedo, L2015 used a constant, 192 193 spatially-invariant value for each land cover class, taken from literature (Nijssen et al. 2001). 194 L2015 did not contain values of fcanopy, leading the VIC model to use a default value of 1.0 everywhere. 195

The MOD-LSP parameter sets obtained monthly phenology from MODIS products: 8day LAI from the MOD15A2H.006 product for the period between 2000-02-18 and 2001-06-26 and the MCD15A2H.006 product for the period 2001-07-04 through the end of 2016 (Myneni et al. 2002); albedo from the "White-Sky Albedo from shortwave broadband" variable in the 1-day MCD43A3.006 product (Schaaf et al. 2002) for the entire period 2000-2016; and  $f_{canopy}$  derived from Normalized Difference Vegetation Index (NDVI) of the 16-day MOD13A1.006 product (Huete et al. 2002) for the entire period 2000-2016. For consistency with LAI, only those albedo

observations corresponding to the 8-day LAI schedule were used. For those 8-day intervals when
the 16-day NDVI data were unavailable, NDVI values were treated as missing data. Phenology
variables were aggregated to the 6 km grid by computing a separate spatial average value for
each land cover class in the grid cell:

207 
$$\bar{x}(c,t) = \frac{1}{N_l(c)} \sum_{k_l=1}^{N_l(c)} x(k_p(k_l),t) \quad , \qquad (1)$$

where *x* is a phenology variable,  $k_l$  is an index of the set of  $N_l(c)$  land cover pixels of class *c* within the cell,  $k_p(k_l)$  is the index of the MODIS pixel containing land cover pixel  $k_l$ , and *t* is the index of the time dimension (at 8-day intervals). For the MOD\_IGBP dataset, land cover pixels were defined on the same grid as phenology pixels (500 m sinusoidal projection), so that each phenology pixel corresponded to exactly one land cover pixel. For the NLCD\_INEGI datasets, the smaller size of land cover pixels (30 m) led to multiple land cover pixels of potentially many different classes corresponding to a single phenology pixel and taking on the same value of *x*.

Pixels that were contaminated by poor retrievals (dead sensor, bad geometry, clouds 215 present) or snow, as indicated by the FparLAI\_QC and FparExtra\_QC variables included in the 216 217 MCD15A2H.006 product, were excluded from the spatial average. In some cases, this led to 218 entire grid cells missing data for one or more acquisition dates, particularly for cells north of  $50^{\circ}$ 219 N latitude in winter (Fig. 3a-c). These missing values were filled by interpolation later in the 220 process flow. Even after filtering out pixels that were flagged for snow, observations immediately before and after the flagged observations often exhibited albedo values much larger 221 than the typical range of variability during the rest of the year, indicating at least partial snow 222 223 coverage. Therefore, after the aggregation step, we further nulled out all observations for which 224 the albedo exceeded four standard deviations above the long-term mean.

The MCD15A2H.006 product did not provide estimates of LAI over pixels classified in the MOD12Q1.005 product as open water, barren, perennial snow/ice, wetlands, urban, or unclassified. With the exception of urban pixels, LAI was set to "missing" in these cases. For urban pixels, LAI was estimated via an empirical LAI-NDVI relationship estimated by sampling pixels of shrubland, grassland, and forest classes surrounding the metropolitan areas of Phoenix, Los Angeles, San Francisco, Portland, and Seattle:

$$LAI = 8 (NDVI - NDVI_{min})^2 , \qquad (2)$$

where  $NDVI_{min} = 0.1$ .

*f<sub>canopy</sub>* was derived from NDVI following Bohn and Vivoni (2016):

234 
$$f_{canopy} = [(NDVI - NDVI_{min})/(NDVI_{max} - NDVI_{min})]^2 , \qquad (3)$$

where 
$$NDVI_{min} = 0.1$$
 and  $NDVI_{max} = 0.8$ .

Gaps (8-day intervals missing data) were filled via the following process (Fig. 3): (1) The time series of each grid cell (Fig. 3d) was separated into its climatological mean and standard deviation (Fig. 3e,f) and standardized anomalies (Fig. 3g), following:

239 
$$x'(c, y, d) = (x(c, y, d) - \mu_x(c, d)) / \sigma_x(c, d) , \qquad (4)$$

Where x' = standardized anomaly, c = land cover class index, y = year, d = day of year, x =phenological variable, and  $\mu_x$  and  $\sigma_x = \text{climatological mean and standard deviation of } x$  for class cand day of year d. (2) All values of  $\mu_x$  and  $\sigma_x$  for which fewer than 5 observations were available on a given day of the year were set to "missing". (3) In the anomaly time series, missing values either on the first or last time steps or for which the nearest valid data points were more than 2 intervals away were set to 0, and remaining gaps in the anomaly time series were linearly

interpolated (Fig. 3h). (4) Temporal gaps in  $\mu_x$  and  $\sigma_x$  were filled by linear interpolation, treating 246 the climatological cycle as periodic (Fig. 3i,j). (5) Anomalies were recombined with  $\mu_x$  and  $\sigma_x$  to 247 assemble the final gap-filled time series (Fig. 3k). (6) Any remaining gaps (cells for which no 248 observations were available at any time) were filled with spatial interpolation from values from 249 250 the same land cover class in neighboring cells, using a Gaussian kernel with  $\sigma = 1$  cell. The final 251 gap-filled data (Fig. 31-n) are therefore estimates of snow- and cloud-free values. In the case of albedo, it is presumed that a land surface model such as VIC would replace the input albedo with 252 253 a simulated snow albedo when snow is present.

254 The requisite repeating annual cycle of monthly phenology was computed from different 255 sets of years for different MOD-LSP parameter sets. For the MOD IGBP.mode parameter set, 256 the gap-filled 8-day timeseries of MODIS phenology spanning 2000-2016 were aggregated to a single year of climatological mean monthly values. For the NLCD\_INEGI.sYYYY (where 257 YYYY = 1992, 2001, or 2011) parameter sets, these climatological mean cycles were computed 258 259 using only those 30-m land cover pixels that had a stable land cover class between the 2001 and 2011 maps, thereby providing a climatological cycle free of the impacts of land cover 260 261 conversion. For the NLCD\_INEGI.YYYY (where YYYY = 2001 or 2011) parameter sets, the 262 annual cycle was taken only from year YYYY of the gap-filled MODIS observations.

Thus, the difference between simulations using 2011 and 2001 parameter sets would show the full impact of land cover conversion and within-class interannual variability in phenology between the years 2001 and 2011, while the difference between simulations using s2011 and s2001 would show the impact of land cover conversion alone. To assess the impact of interannual variability of phenology alone, an additional set of 17-year monthly time series was prepared for all MOD-LSP parameter sets via aggregating from the gap-filled 8-day records of

LAI, *f<sub>canopy</sub>*, and albedo. For the NLCD\_INEGI.2001 and NLCD\_INEGI.2011 parameter sets, these time series include the impact of land cover conversion on the time series of phenology (e.g., forest-shrub conversion due to wildfire or logging); however the classification of the pixels in these cases will be that of either year 2001 or 2011.

273 2.3.4. Non-MODIS Surface Properties

All parameter sets obtained grid cell elevations from the USGS GTOPO30 digital 274 elevation model (Gesch et al. 1999). For the MOD-LSP datasets, GTOPO30 was aggregated 275 directly to 0.0625° resolution, but L2015 was sub-sampled from earlier 0.125° (12 km) spatial 276 resolution datasets; thus their land masks differ along the coastlines. All VIC parameter datasets 277 used the soil properties of L2015, which were obtained from the FAO-UNESCO Digital Soil 278 Map of the World (FAO/UNESCO 1998). Similarly, all VIC parameter datasets used the 279 280 conceptual soil parameter values (e.g., *binfilt*, *Dsmax*, and soil layer thicknesses) of L2015, which were derived via calibration in previous studies (Maurer et al. 2002; Zhu and Lettenmaier 2007). 281 282 For all VIC parameter datasets, time- and space-invariant surface properties for each land 283 cover class (e.g., presence of overstory, stomatal and architectural resistance parameters) were 284 taken from the L2015 dataset, which in turn obtained these values from literature (Nijssen et al. 285 2001). Because the new parameter datasets used different land cover classification schemes than 286 the L2015 dataset, these time- and space-invariant parameters were mapped to the most 287 appropriate class of the UMD-NLDAS classification as shown in Tables 4 and 5. Because the 288 L2015 dataset did not contain open water, perennial ice/snow, or urban classes, properties for 289 these classes in the MOD\_IGBP and NLCD-INEGI datasets were taken from the grassland class 290 of the L2015 dataset. However, the grassland properties only applied to the vegetated portions of 291 these classes, which were prescribed by MODIS observations of LAI and *f*canopy. Furthermore,

the properties of the open water class can be overridden with those of water when the VIC lakemodel is turned on.

#### 294 2.4. Evaluation Datasets

To evaluate the values of  $f_{canopy}$  in the new parameter datasets, two independent datasets were used: the NLCD US Forest Service Percent Tree Canopy Cartographic product (Coulston et al. 2012) (NLCD-Forest hereafter) and the NLCD Shrubland product (Xian et al. 2015) (NLCD-Shrub hereafter). These datasets were aggregated from 30-m to 0.0625 ° resolution over the NLCD\_INEGI.2011 land cover classification in a similar fashion to the MODIS observations.

300 2.5. Hydrologic Model

The VIC model formulation is described in detail elsewhere (Liang et al. 1994; Hamman 301 et al. 2018). As of release 4.2, the VIC model optionally can read the annual cycle of fcanopy from 302 the parameter file, and when  $f_{canopy} < 1.0$ , soil evaporation (*Esoil*) will be computed in the gaps 303 between plants (Bohn and Vivoni 2016); otherwise, as in prior releases of the model, *f<sub>canopy</sub>* is 304 assumed to equal 1.0 for all land cover classes except the explicit bare soil class. As of release 305 5.0, in image mode, spatially explicit annual cyles of LAI and albedo must be specified in the 306 307 parameter file. For albedo, the values from the parameter file are used under snow-free conditions, but the simulated snow albedo is used when snow is present. As of release 4.2, the 308 VIC model can optionally read time series of phenology as additional forcing variables, but as 309 310 such must be specified at the same time step as the meteorological forcings, which are typically hourly (Bohn and Vivoni 2016). Although the VIC model contains a lake component (Bowling 311 and Lettenmaier 2010), activating it requires additional parameters that are not included in the 312 MOD-LSP parameter sets. 313

314 2.6. Public Access

315	All parameter sets discussed herein are stored in NetCDF files, formatted for input to the
316	Variable Infiltration Capacity (VIC) model (Liang et al. 1994), version 5.0 and later (Hamman et
317	al. 2018), in image mode. The parameter sets are available for download at Zenodo (Bohn and
318	Vivoni 2019a). The L2015 parameter set was converted from ascii to VIC-5-compliant NetCDF
319	format by the "tonic" tool (https://github.com/UW-Hydro/tonic/releases/0.2). The MOD-LSP
320	parameter sets were created via Python scripts (that use the xarray package (Hoyer and Hamman
321	2017)) from the VIC_Landcover_MODIS_NLCD_INEGI project (Bohn 2019b) archived on
322	GitHub (https://github.com/tbohn/VIC_Landcover_MODIS_NLCD_INEGI/releases/tag/v1.4).
323	The NLCD_INEGI harmonized US-Mexico land cover classifications for years 1992/3, 2001/2,
324	and 2011 are available for download at Zenodo (Bohn and Vivoni 2019b). The scripts used to
325	create the NLCD_INEGI harmonized land cover classifications are archived on GitHub
326	(https://github.com/tbohn/NLCD_INEGI/releases/tag/v1.5) (Bohn 2019a).
327	
328	3. Results and Discussion
329	3.1. Forest and Shrubland Canopy Fraction Evaluation
330	The MOD-LSP estimates of $f_{canopy}$ performed better, relative to the NLCD-Shrub and
331	NLCD-Forest products, than L2015 (Table 6; Fig. 4). July average values of $f_{canopy}$ from
332	NLCD_INEGI.s2011 tended to underestimate the NLCD-Shrub values and overestimate NLCD-
333	forest values by factors of 0.70 and 1.24, respectively, yielding RMSE values of 0.09 and 0.21,

- respectively. However, the L2015 parameter set, which implicitly assumed  $f_{canopy} = 1.0$  in all
- vegetated classes, grossly overestimated the NLCD-Shrub and NLCD-Forest values by factors of
- 336 7.24 and 2.28, and yielded RMSE values of 0.87 and 0.61.

337	It should be noted that Eqn (3) neglects impacts of non-linearity of spatial aggregation of
338	heterogeneous NDVI and impacts of soil brightness, soil wetness, and canopy shading (Jiang et
339	al. 2006). Thus, there will be local biases due to local soil and vegetation properties; furthermore,
340	some of the seasonal cycle in $f_{canopy}$ estimated via Eqn (3) are due to seasonal changes in solar
341	angle. Thus, this formula could stand to be refined. Nevertheless, Bohn and Vivoni (2016) found
342	that VIC simulations using $f_{canopy}$ values derived from MODIS NDVI via Eqn (3) resulted in
343	substantial improvements in simulated ET at 60+ shrubland and forest AMERIFLUX (Baldocchi
344	et al. 2001) eddy covariance tower sites.
345	3.2. Urban Phenology Evaluation
346	A comprehensive evaluation of urban LAI and $f_{canopy}$ from MOD-LSP has not been
347	undertaken. However, over Santa Barbara, CA, July values of LAI and <i>f</i> <sub>canopy</sub> compare favorably
348	to 71 randomly-sampled field observations (Alonzo et al. 2015). Observed LAI values had a
349	mean of 1.0035 and standard error of 0.1097. The NLCD_INEGI.s2011 LAI values, averaged
350	over all urban classes in the grid cells containing Santa Barbara, had a mean of 0.8843. Alonzo et
351	al. (2015) reported canopy coverage values of 0.20 (from the UFORE model) and 0.254
352	(estimated from high-resolution imagery by the city of Santa Barbara). The NLCD_INEGI.s2011
353	July urban $f_{canopy}$ values had a mean of 0.2256.
354	3.3. Comparison with Prior Parameter Set – Land Cover Distributions
355	The L2015, MOD_IGBP, and NLCD_INEGI.2011 land cover classifications generally
356	agree on the geographic distributions of the major land cover categories, although there are
357	notable differences (Fig. 5). For the "forest" category (the sum of all forest and wooded
358	grassland or savanna classes), all three land cover datasets show high coverage in eastern Canada

and northeastern United states, the southeastern United States, southern Mexico, the mountains

west of 100 °W longitude, and the Pacific Northwest coast (Fig. 5a-c). However, L2015 has 360 higher forest coverage around the fringes of these regions than MOD IGBP, while 361 362 NLCD\_INEGI has lower coverage. This might be due in part to the spatial resolutions of the underlying land cover products (1 km for UMD, 500 m for MOD12Q1.005, and 30 m for NLCD 363 and INEGI). In particular, the NLCD products do not include classes for mixes of trees and grass 364 365 (e.g., savanna), likely due to the difficulty of identifying such a mixture from an individual 30-m pixel. For shrublands (Fig. 5d-f) and grasslands (Fig. 5g-i), the three land cover classifications 366 agree on the general locations of these classes (arid western North America) but differ on the 367 boundary between shrubland and grassland. L2015 is more similar to NLCD\_INEGI in its 368 partitioning between shrublands and grasslands, with shrublands extending into the northwestern 369 United states, while in MOD12Q1.005 shrublands are confined to the southwestern United States 370 and western Mexico. The three products again agree on the general distribution of agriculture 371 (Fig. 5j-l), but MOD\_IGBP has the greatest agricultural coverage in most regions of the US and 372 373 Canada, and NLCD\_INEGI has the greatest coverage in Texas and Mexico. 3.4. Comparison with Prior Parameter Set - Phenology 374 The improvements of MOD-LSP phenology over that of prior studies are evident when 375 comparing maps over North America. For the "mixed forest" class (Fig. 6), the July LAI of 376 L2015 (Fig. 6a) contains regions of homogeneous values with abrupt boundaries, while 377 378 MOD\_IGBP.mode (Fig. 6d) exhibits a more physically reasonable distribution. The L2015 July fcanopy (Fig. 6b) is 1.0 everywhere, while that of MOD\_IGBP.mode (Fig. 6e) varies from 0.5 to 379 1.0 and shows a similar spatial distribution to LAI. Similarly, the L2015 July albedo (Fig. 6c) is 380 381 0.18 everywhere, while that of MOD\_IGBP.mode (Fig. 6f) ranges from 0.10 to 0.18. Similar

problems are evident in the "cropland" class of L2015, relative to that of MOD\_IGBP.mode(Fig. 7).

384	MOD-LSP and L2015 parameter sets also differ in total "scene" phenology (the area-
385	weighted average phenology across all land cover classes) (Fig. 8). In January, the "scene" LAI
386	of L2015 (Fig. 8a) is substantially higher than that of MOD_IGBP.mode (Fig. 8g), particularly in
387	the northern forests. This could be due to both (1) L2015 holding LAI constant year-round for
388	evergreen forest and (2) possible residual impacts from clouds and snow that were not
389	completely accounted for in the MOD-LSP processing. However, in July, the LAI of L2015 (Fig.
390	8d) and MOD_IGBP.mode (Fig. 8j) look more similar. <i>f</i> canopy remains 1.0 year-round in all
391	vegetated cells in L2015 (Fig. 8b,e) but displays a seasonal cycle in most of the domain in
392	MOD_IGBP.mode (Fig. 8h,k). For albedo, spatial variation is evident in L2015, ranging from
393	0.12 to 0.20, but there is no temporal variation except over croplands (Fig. 8c,f). The
394	MOD_IGBP.mode albedo exhibits much larger spatial variability than L2015, ranging from 0.10
395	to 0.30 (Fig. 8i,l).

396 3.5. Impacts on Hydrology

The aforementioned differences in land cover distributions and phenology between the L2015 and MOD-LSP parameter sets (Figs 5-8) impact water and energy flux partitioning in hydrology simulations. Holding  $f_{canopy}$  constant at 1.0 (Fig. 9), the differences in mean annual ET (and Q) between simulations using the MOD\_IGBP and L2015 parameter sets are positively (negatively) correlated with the difference in mean annual LAI between the two parameter sets. In the warmer and wetter regions of the domain (south and east of the dashed lines), ET and Q exhibit sensitivities to differences in LAI of about +/-200 mm y<sup>-1</sup> per unit change in LAI.

404 However, the differences diminish in the cooler and drier portions of the domain (to the north405 and west of the dashed lines).

406 Similarly, allowing  $f_{canopy}$  to vary (Fig. 10) results in changes of about -/+50 mm y<sup>-1</sup> per 407 unit change in *f<sub>canopy</sub>* in the warmer and wetter portions of the domain, due to reduced canopy evaporation being outweighed by the increased soil evaporation and transpiration from increased 408 409 throughfall (Fig. 10). Again, sensitivities to changes in  $f_{canopy}$  diminish as climate becomes drier, but where annual P falls primarily as snow (e.g., the western mountains), the sensitivities of ET 410 411 and Q to changes in *f<sub>canopy</sub>* change sign, due to reduced snow storage in the forest canopy, where aerodynamic resistance is lower (and turbulent fluxes higher) than at the surface of the ground 412 413 snow pack. The total difference between MOD IGBP with varying  $f_{canopy}$  and L2015 is dominated by the difference due to LAI, but the  $f_{canopy}$  impacts reinforce the LAI impacts in the 414 western mountains. 415

416 3.6. Land Cover Change

The NLCD\_INEGI products show substantial land cover change between 2001 and 2011 417 (Fig. 11). Forests experienced substantial losses in the southeastern United States, the Pacific 418 Northwest, and much of Mexico (Fig. 11a,b). Shrublands and grasslands expanded in the 419 southeastern US and Pacific Northwest, but shrank in northern Mexico (Fig. 11g,h). Agriculture 420 expanded substantially in Mexico (Fig. 11m,n) and cities expanded everywhere (Fig. 11s,t). At 421 422 smaller scales (boxes 1-4 in Fig. 11b,h,n,t), the NCLD\_INEGI products show the loss of forest to shrub/grassland in response to drought and fire in the Rocky Mountains (box 1; Breshears et al. 423 2005); a mix of forest-shrub, crop-forest, and crop-urban conversion around Lake Michigan (box 424 425 2); forest-shrub, forest-crop, and shrub-crop conversion in northwestern Mexico due to pasture

clearing and agricultural expansion (box 3; Bohn et al. 2018a); and forest-shrub and forest-urban
conversion around Atlanta (box 4).

428	These changes in land cover, coupled with interannual variations in phenology (via the
429	NLCD_INEGI.2001.2001_2001 and NLCD_INEGI.2011.2011_2011 parameter sets), led to
430	changes in hydrologic fluxes (Fig. 12). Forest-shrub conversion in the Rocky Mountains (box 1)
431	and forest-agriculture conversion in northwestern Mexico (box 3) were accompanied by
432	reductions in LAI and <i>f</i> <sub>canopy</sub> , leading to reduced ET and increased Q. However, crop-urban
433	conversion around Lake Michigan (box 2) and forest-urban conversion around Atlanta (box 4)
434	did not lead to reductions in LAI or <i>f</i> <sub>canopy</sub> and had minimal impacts on hydrology. Furthermore,
435	interannual fluctuations in LAI and $f_{canopy}$ in eastern Texas (box 5) led to substantial changes in
436	hydrology without being accompanied by substantial land cover change.

437

## 438 4. Parameter Sets, File Names, and Configuration for Running VIC

There are three types of files included in the MOD-LSP project (Table 7): (1) VIC 439 440 parameter files, which contain land cover class area fractions ("Cv") in all grid cells, spatially-441 varying annual cycles of monthly phenology variables ("LAI", "fcanopy", "albedo") for all classes in all grid cells, soil properties, and vegetation structural properties that are invariant in 442 443 time and space; (2) "veg hist" compressed tar files, each containing 17 yearly files of monthly time-varying phenology variables; and (3) a template ("global param.template") for the global 444 445 parameter file, which specifies the names, locations, and contents of all input and output files, and simulation parameters such as model time step, start and end dates, and model physics 446 options. 447

To create a global parameter file from the template, the user should make a copy of the template
and edit it. Lines beginning with "#" are comment statements that are ignored by the VIC model.
All words enclosed in "<>" are placeholders that need to be replaced by the desired values (the
"<>" must be removed). All options in the global parameter file are explained in the VIC model
documentation (<u>https://vic.readthedocs.io</u>) but the most relevant options are explained in Table 8.
Notable variables in the parameter and veg\_hist files are described in Table 9.

The veg\_hist files are optional; if omitted, the VIC model will use the repeating annual 454 cycle of phenology from the parameter file. These files must contain data at the same time step 455 length as the meteorological forcings. To use the veg\_hist files requires (1) disaggregating these 456 files of monthly values to an hourly time step (assuming meteorology is hourly) using the scripts 457 *disagg\_veghist\_monthly2hourly\_nc.py* and *wrap\_disagg\_veghist\_monthly2hourly\_nc.pl* from the 458 VIC\_Landcover\_MODIS\_NLCD\_INEGI GitHub project 459 (https://github.com/tbohn/VIC\_Landcover\_MODIS\_NLCD\_INEGI/releases/tag/v1.4) and (2) 460 461 setting the appropriate options in the global parameter file (Table 8). Instructions for this

462 procedure can be found in the GitHub archive.

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## 601 Tables

602	Table 1. Descri	ptions of spatial	datasets used in this study.
		perono or operent	

Туре	Abbreviation	Dataset	Reference	Satellite	Acquisition	Temporal	Resolution
				Source	Date Range	Interval	
Land Cover Classification	UMD	University of Maryland Land Cover	(Hansen et al. 2000)	AVHRR	1981-1994	Single map	1 km
	MOD12Q1.051	MODIS Collection 5 Global Land Cover	(Friedl et al. 2010)	MODIS	2001-2013	Annual	500 m
	NLCD	National Land Cover Database	(Homer et al. 2007; Fry et al. 2009; Homer et al. 2015)	LANDSAT	1992-2011	Years 1992, 2001, 2006, 2011	30 m
	INEGI	Uso del Suelo y Vegetación	(INEGI 2014)	LANDSAT	1992-2011	Years 1993, 2002, 2007, 2011	n/a (imagery delineated by hand into polygons)
Static Surface Properties	GTOPO30	Global 30 Arc- Second Digital Elevation Model	(Gesch et al. 1999)	n/a	n/a	Single map	30" (1 km)
	NLCD-Forest	National Land Cover Database USFS Tree Cover, Cartographic	(Coulston et al. 2012)	LANDSAT	2008-2009	Single map	30 m
	NLCD-Shrub	National Land Cover Database Shrub Cover	(Xian et al. 2015)	WorldView-2, LANDSAT	2013	Single map	30 m
	FAO	FAO-UNESCO Digital Soil Map of the World	(FAO/UNESCO 1998)	n/a	n/a	Single map	5' (8 km)
Time-Varying Surface	B2002	Leaf Area Index	(Myneni et al. 1997; Buermann et al. 2002)	AVHRR	1981-1994	10 day	5' (8 km)
Properties	MCD15A2H.006	Leaf Area Index	(Myneni et al. 2002)	MODIS	2000-2016	8 day	500 m
	MOD13A1.006	Normalized Difference Vegetation Index	(Huete et al. 2002)	MODIS	2000-2016	16 day	500 m

	MCD43A3.006	Albedo	(Schaaf et al. 2002)	MODIS	2000-2016	1 day	500 m
.02							

Land Cover Code	Domain	Soil	Land Cover Classification		F	Phenology		Comment	
(lc_source.lc_year)		Properties	Source	Temporal Coverage	Classification Scheme	Source	Seasonal cycle	Time Series	
L2015	CONUS_MX	FAO	UMD	Single snapshot representing 1981-1994	UMD, with open water and urban replaced by bare soil	LAI: AVHRR <i>f<sub>canopy</sub></i> : n/a Albedo: n/a	Single cycle from 1992- 1993	n/a	Phenology derived by spatial interpolation of values from < 50 pixels in each class
MOD_IGBP.mode	CONUS_MX	FAO	MODIS MOD12Q1.051	For each pixel, mode of class over 2001-2013	IGBP	LAI: MCD15A2H.006 <i>f</i> <sub>canopy</sub> : MOD13A1.006	Monthly climatology 2000-2016	Monthly, 2000- 2016	
NLCD_INEGI.2011	USMX	FAO	NLCD over	2011	NLCD 2011	Albedo:	2011	n/a	
NLCD_INEGI.2001	USMX	FAO	US; INEGI over Mexico	2001 over US; 2002 over Mexico	NLCD 2011	MCD43A3.006 (Taken from "White-Sky	2001	n/a	
NLCD_INEGI.s2011	USMX	FAO		2011	NLCD 2011	Albedo for shortwave broadband")	Monthly climatology 2000-2016	Monthly, 2000- 2016	
NLCD_INEGI.s2001	USMX	FAO		2001 over US; 2002 over Mexico	NLCD 2011		from pixels with stable class	Monthly, 2000- 2016	
NLCD_INEGI.s1992	USMX	FAO		1992 over US; 1993 over Mexico	NLCD 2011		between 2001 and 2011	n/a	Over US, NLCD 1992- 2001 retrofit change product applied to 2001, preserving relative proportions of shrub and grass relative to their total and of crop and pasture relative to their total

**Table 2.** Description of various versions of VIC input parameters. All VIC parameters have 0.0625 degree (6 km) spatial resolution.

Table 3. Primary INEGI codes and descriptions, mappings to NLCD classes. <u>Yellow</u> indicates
 codes present in 1993 only. <u>Green</u> indicates codes not present in 1993.

Code	Description	NLCD ClassName <sup>1</sup>	NLCD ClassNum <sup>1</sup>	Comment
[R]	Area Agricola; Riego suspendido	Cultivated Crops	82	Present in 1993 only
ACUI	Acuicola	Cultivated Crops	82	Not present in 1993
ADV	Desprovisto de veg.	Barren Land	31	Not present in 1993
AH	Asentamientos humanos	Developed, low intensity	22	"Human Settlements"; Not present in 1993
BA	Bosque de oyamel	Needleleaf Forest	42	"Fir forest"
BB	Bosque cedro	Needleleaf Forest	42	"Cedar forest"
BC	Bosque cultivado	Broadleaf Forest	41	"Cultivated forest"; assumed to be orchards
BG	Bosque de galeria	Broadleaf Forest	41	"Riparian forest"
BI	<b>Bosque inducido</b>	<b>Broadleaf Forest</b>	<mark>41</mark>	Not present in 1993
BJ	Bosque de tascate	Needleleaf Forest	42	"Juniper forest"
BM	Bosque mesofilo de montana	Broadleaf Forest	41	"Cloud forest"; assumed predominantly broadleaf
BP	Bosque de pino	Needleleaf Forest	42	"Pine forest"
BPQ	Bosque de pino- encino	Mixed Forest	43	"Pine-Oak forest"
BQ	Bosque de encino	Broadleaf Forest	41	"Oak forest"
BQP	Bosque de Encino- pino	Mixed Forest	43	"Oak-Pine forest"
BS	Bosque de ayarin	Needleleaf Forest	42	"Spruce forest"
<mark>BW</mark>	Bosque de bajo abierto	Broadleaf Forest	<mark>41</mark>	Present in 1993 only
DV	Sin veg. aparente	Barren Land	31	
H2O	Cuerpo de agua	Open Water	11	
HA	humedad	Cultivated Crops	82	Rainfed, Annual
HAP	humedad	Cultivated Crops	82	Rainfed, Annual, permanent
HAS	humedad	Cultivated Crops	82	Rainfed, Annual, semi-permanent
HP	humedad	Cultivated Crops	82	Rainfed, Permanent
HS	humedad	Cultivated Crops	82	Rainfed, Semi-permanent
HSP	humedad	Cultivated Crops	82	Rainfed, Semi-permanent/ permanent
MC	Matorral crasicuale	Shrub/Scrub	52	
MDM	Matorral desertico microfilo	Shrub/Scrub	52	
MDR	Matorral desertico rosetofilo	Shrub/Scrub	52	
MET	Matorral espinoso tamauli-peco	Shrub/Scrub	52	
<mark>MJ</mark>	Matorral de coniferas	Shrub/Scrub	<mark>52</mark>	Present in 1993 only
MK	mezquital	Shrub/Scrub	52	
MKE	Mezquital tropical	Shrub/Scrub	<mark>52</mark>	Not present in 1993
MKX	Mezquital xerofilo	Shrub/Scrub	52	Not present in 1993
ML	chaparral	Shrub/Scrub	52	
MRC	Matorral rosetofilo	Shrub/Scrub	52	
	costero			
MSC	Matorral sarcocuale	Shrub/Scrub	52	

MSCC	Matorral sarco-	Shrub/Scrub	52	
	crasicuale			
MSM	Matorral sub-	Shrub/Scrub	52	
MSN	Matorral sarco-	Shrub/Scrub	52	
	crasicuale de neblina	~		
MST	Matorral subtropical	Shrub/Scrub	52	
MU	huizachal	Shrub/Scrub	<u>52</u>	Present in 1993 only
P/E	Pais extranjero	(nodata)	-1	Outside the boundaries of Mexico
PA D C	Pastizal - huizachal	Grassland/Herbaceous	<u>71</u>	Present in 1993 only
PC	Pastizal cultivado	Pasture/Hay	81	
PH	Pastizal halofilo	Grassland/Herbaceous	71	Assumed to be natural salt- tolerant grassland
PI	Pastizal inducido	Pasture/Hay	81	
PN	Pastizal natural	Grassland/Herbaceous	71	
PT	Veg. de Peten	Woody Wetland	<mark>90</mark>	Not present in 1993 (was "manglar")
PY	Pastizal gipsofilo	Grassland/Herbaceous	71	
RA	riego	Cultivated Crops	82	Irrigated, Annual
RAP	riego	Cultivated Crops	82	Irrigated, Annual/ permanent
RAS	riego	Cultivated Crops	82	Irrigated, Annual/ semi-permanent
RP	riego	Cultivated Crops	82	Irrigated, Permanent
RS	riego	Cultivated Crops	82	Irrigated, Semi-permanent
RSP	riego	Cultivated Crops	82	Irrigated, Semi-permanent/
				permanent
SAP	Selva alta perenni- folia	Broadleaf Forest	41	
SAQ	Selva alta sub- perenni-folia	Broadleaf Forest	41	
SBC	Selva baja caducifolia	Broadleaf Forest	41	
SBK	Selva baja espinosa	Broadleaf Forest	41	
SBP	Selva baja perreni- folia	Broadleaf Forest	41	
SBQ	Selva baja sub- perreni-folia	Broadleaf Forest	41	
SBQP	Selva baja sub- perreni-folia	Broadleaf Forest	41	
SBS	Selva baja subcaduci- folia	Broadleaf Forest	41	
SG	Selva de galeria	Broadleaf Forest	41	"Riparian forest"
SMC	Selva mediana	Broadleaf Forest	41	
SMP	Selva mediana	Broadleaf Forest	41	
SMQ	Selva mediana sub-	Broadleaf Forest	41	
SMS	Selva mediana subcaduci-folia	Broadleaf Forest	41	
ТА	Temporal	Cultivated Crops	82	Rainfed, Annual
ТАР	Temporal	Cultivated Crops	82	Rainfed, Annual/ permanent
TAS	Temporal	Cultivated Crops	82	Rainfed, Annual/ permanent
TP	Temporal	Cultivated Crops	82	Rainfed, permanent
TS	Temporal	Cultivated Crops	82	Rainfed, permanent

TSP	Temporal	Cultivated Crops	82	Rainfed, permanent
VA	Popal	Emergent Herbaceous	95	Emergent veg – herbaceous
		Wetland		(lily/water hyacinth)
VD	Veg. de desiertos	Barren Land	31	VD areas tend to be classified as
	arenosos			"bare" in MODIS and NLCD
				classifications
VG	Veg. de galeria	Woody Wetland	90	"Riparian vegetation"; assume that
				there are shrubs and trees
VH	Veg. halofila	Emergent Herbaceous	95	Salt marsh (halophyllic veg)
		Wetland		
VHH	Veg. halofila hidrofila	Emergent Herbaceous Wetland	<u>95</u>	Not present in 1993
VM	Manglar	Woody Wetland	90	"Mangrove swamp"
<b>VP</b>	Palmar	Broadleaf Forest	<mark>41</mark>	"Palm"; Present in 1993 only
VPI	Palmar Inducido	<b>Broadleaf Forest</b>	<mark>41</mark>	"Palm"; Not present in 1993
VPN	Palmar natural	<b>Broadleaf Forest</b>	<mark>41</mark>	"Palm"; Not present in 1993
VS	Sabana	Broadleaf Forest	41	"savanna"
VSI	Sabanoide	Broadleaf Forest	<mark>41</mark>	Assumed similar to savanna; Not present in 1993
VT	Tular	Emergent Herbaceous Wetland	95	Cattails, reeds, etc
VU	Veg. de las dunas	Shrub/Scrub	52	
VIII	Dradara da alta	Creaseland/Harbaasaya	71	
V VV	montana	Grassland/Herbaceous	/1	
VY	Veg. gipsofila	Shrub/Scrub	52	
ZU	Zona urbana	Developed, High	24	
		intensity		

<sup>1</sup>The NLCD classes of "Deciduous Forest" and "Evergreen Forest" (codes 41 and 42) were interpreted to correspond to broadleaf and needleleaf forest, respectively, in Mexico. Over the CONUS\_MX domain, Deciduous Needleleaf

610 Forest is extremely rare, so that the NLCD Evergreen Forest class essentially denotes Needleleaf Forest. Similarly,

611 in the UMD-NLDAS dataset, the time-invariant structural parameters for Deciduous Broadleaf Forest were the same

as those for Evergreen Broadleaf Forest, so that our interpretation did not adversely impact parameter values.

613

- **Table 4.** Mapping between MOD-IGBP classes and UMD-NLDAS classes from which structural
- 616 parameters were taken.

MOD-IGBP		UMD-NLDAS	
Code	Names	Code Name	
0	Open Water	10	Grassland
1	Evergreen Needleleaf	1	Evergreen Needleleaf Forest
	Forest		
2	Evergreen Broadleaf Forest	3	Evergreen Broadleaf Forest
3	Deciduous Needleleaf	2	Deciduous Needleleaf
	Forest		Forest
4	Deciduous Broadleaf Forest	4	Deciduous Broadleaf Forest
5	Mixed Forest	5	Mixed Forest
6	Closed Shrubland	8	Closed Shrubland
7	Open Shrubland	9	Open Shrubland
8	Woody Savanna	6	Woodland
9	Savanna	7	Savanna
10	Grassland	10	Grassland
11	Wetland	10	Grassland
12	Cropland	11	Cropland
13	Urban	10	Grassland
14	Crop-Natural Mosaic	11	Cropland
15	Perennial Ice/Snow	10	Grassland
16	Barren	10	Grassland

Table 5. Mapping between NLCD-INEGI classes and UMD-NLDAS classes from whichstructural parameters were taken.

NLCD-INEGI		UMD-NLDAS	
Code	Names	Code	Name
11	Open Water	10	Grassland
12	Perennial Ice/Snow	10	Grassland
21	Developed, Open Space	10	Grassland
22	Developed, Low Intensity	10	Grassland
23	Developed, Medium	10	Grassland
	Intensity		
24	Developed, High Intensity	10	Grassland
31	Barren Land	10	Grassland
41	Deciduous Forest <sup>1</sup>	4	Deciduous Broadleaf Forest
42	Evergreen Forest <sup>1</sup>	1	Evergreen Needleleaf Forest
43	Mixed Forest	5	Mixed Forest
52	Shrub/Scrub	9	Open Shrubland
71	Grassland/Herbaceous	10	Grassland
81	Pasture and Hay	11	Cropland
82	Cultivated Crops	11	Cropland
90	Woody Wetlands	10	Grassland
95	Emergent Herbaceous	10	Grassland
	Wetlands		

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<sup>1</sup>The NLCD classes of "Deciduous Forest" and "Evergreen Forest" (codes 41 and 42) were interpreted to correspond

to broadleaf and needleleaf forest, respectively, in Mexico. Over the CONUS\_MX domain, Deciduous Needleleaf

624 Forest is extremely rare, so that the NLCD Evergreen Forest class essentially denotes Needleleaf Forest. Similarly,

in the UMD-NLDAS dataset, the time-invariant structural parameters for Deciduous Broadleaf Forest were the same
 as those for Evergreen Broadleaf Forest, so that our interpretation did not adversely impact parameter values.

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- **Table 6.** Statistics of L2015 and MOD\_LSP (NLCD\_INEGI.2011) July *f*<sub>canopy</sub> values relative to
- NLCD-Shrub (Xian et al. 2015) and NLCD-Forest (Coulston et al. 2012) products. The subscript
  "ref" refers to NLCD-Shrub and NLCD-Forest.

Lar	nd Cover Category	<b>Parameter Set</b>	Statistics			
			μ	μ / μ <sub>ref</sub>	Pearson r	RMSE
Shr	ub	L2015	1.0000	7.2368	0.0000	0.8658
		NLCD_INEGI.2011	0.0962	0.6964	0.3643	0.0940
Fore	est	L2015	1.0000	2.2810	0.0000	0.6120
		NLCD_INEGI.2011	0.5458	1.2449	0.6893	0.2064
633						

#### **Table 7.** Files included in this dataset.

Name	Description		
global_param.template	Template for global parameter file for VIC model		
params.domain.lc_source.lc_year.phen_years.nc	<ul> <li>VIC soil and vegetation parameters. In these files, phenology data (LAI, <i>f<sub>canopy</sub></i>, albedo) for each class in each grid cell consist of a single set of 12 climatological monthly mean values. File naming convention is as follows:</li> <li><i>domain</i> = domain name ("CONUS_MX" or "USMX")</li> <li><i>lc_source</i> = land cover source name ("L2015", "MOD_IGBP", or "NLCD_INEGI")</li> <li><i>lc_year</i> = code describing the year(s) of acquisition from which pixels are assigned classes; can be one of ("mode", <i>year</i>, or "s"<i>year</i>) where: <ul> <li><i>mode</i> = class is the most common class from all maps over the period 2000-2013</li> <li><i>year</i> = class comes from the land cover map for year <i>year</i></li> <li>"s"<i>year</i> = same as <i>year</i>, but the "s" signifies that the phenology data were only aggregated over pixels for which land cover class was stable (did not change) between 2001 and 2011. Pixels for which land cover class changed were assigned phenology values via spatial interpolation from stable neighbors of the class to which the pixel belonged in the given <i>lc_year</i>.</li> </ul> </li> <li><i>phen_years</i> = years of MODIS products from which phenology values are derived, in the format <i>startyear_endyear</i>.</li> </ul>		
veg_hist.domain.lc_source.lc_year.phen_years.monthly.tgz	<ul> <li>Gzipped tar file of directory containing yearly files (one file per year) of time-varying monthly phenology. Individual files follow the naming convention:</li> <li>veg_hist.domain.lc_source.lc_year.phen_years.monthly.year.nc</li> <li>where <ul> <li>domain.lc_source.lc_year.phen_years = same naming convention as for the params files</li> <li>year = year of phenology to which this file pertains</li> </ul> </li> <li>Note: these files must be converted from monthly values to sub-daily values (same time step as the meteorological forcings) via the following scripts, available at https://github.com/tbohn/VIC_Landcover_MODIS_NLCD_INEGI/releases/tag/v1.4:</li> <li>disagg_veghist_monthly2hourly_nc.pp</li> <li>wrap_disagg_veghist_monthly2hourly_nc.pl</li> </ul>		

Option	Description	Default
<domain_file></domain_file>	Path/name of the domain file. The land mask in	None.
	the domain file determines the spatial domain	
	of the simulation. The spatial dimensions	
	(resolution, numbers of rows and columns,	
	latitudes and longitudes of grid cell centers)	
	must match those in the desired VIC parameter	
	file exactly. Domain files covering the	
	CONUS_MX and USMX domains are	
	available on Zenodo at	
	( <u>https://zenodo.org/record/2564019</u> )(Bohn et	
	al. 2018b).	
<param_file></param_file>	Path/name of the desired parameter file.	None.
<lai_src>,</lai_src>	Options to tell the VIC model where to obtain	<fcan_src>:</fcan_src>
<fcan_src>,</fcan_src>	phenology variables from. To use the repeating	FROM_DEFAULT
<alb_src></alb_src>	annual cycles of monthly values stored in the	All others:
	parameter file, these should be set to	FROM_VEGPARAM
	FROM_VEGPARAM. To use the monthly	
	time series of phenology from the veg_hist	
	files, these should be set to FROM_VEGHIST	
	(see below for instructions on preparing the	
	veg_hist files). Because the L2015 parameter	
	file does not contain <i>f</i> <sub>canopy</sub> values and has no	
	associated veg_hist file, <fcan_src> must</fcan_src>	
	be set to FROM_DEFAULT when using the	
	L2015 parameter file (which instructs VIC to	
	use a value of 1.0 everywhere).	
<forcing_dir></forcing_dir>	Path of the directory containing meteorological	None.
	forcing files.	
<forcing_pfx></forcing_pfx>	Prefix of meteorological forcing files. Forcing	None.
	files are assumed to be named as	
	$forcing\_pfx.YYYY.nc$ , where $YYYY =$ the	
	year covered by the file.	
<pre><veghist_dir></veghist_dir></pre>	Path of the directory containing veg_hist files.	Use repeating cycle
		trom parameter file.
<pre><veghist_pfx></veghist_pfx></pre>	Prefix of veg_hist files. Veg_hist files are	Use repeating cycle
	assumed to be named as <i>veghist_pfx</i> . YYYY.nc,	from parameter file.
	where $Y Y Y Y =$ the year covered by the file.	

**Table 8.** Explanation of placeholders in global\_param.template file.

Name	Definition				
veg_descr	Land cover class name				
veg_class	Class ID code				
mask	Land mask. The MOD-LSP parameter files have a slightly different mask than the				
	L2015 file, primarily along coastlines, due to the higher-resolution processing of				
	the DEM in the MOD-LSP processing.				
run_cell	Mask determining which cells to run in the VIC simulation. Run_cell is currently				
	set to equal the L2015 mask in the MOD_IGBP parameter file, to allow it to be				
	used with the L2015 meteorological forcings. Similarly, the NLCD_INEGI				
	parameter files set run_cell to the intersection of the L2015 and USMX land				
	masks. To use these parameter files with a different set of meteorological forcings,				
	run_cell must be modified to be consistent with the land mask of those forcings.				
LAI	Monthly Leaf Area Index				
fcanopy	Monthly <i>fcanopy</i>				
albedo	Monthly albedo				

640	Table 9. Definitions of sele	ected variables in the pa	arams and veg_hist files.
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## 643 Figures



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- **Fig. 1.** Example process flow for LAI, showing spatial aggregation of MODIS LAI over a land
- 646 cover classification to generate spatial average values for each land cover class at 6 km
- 647 resolution.



- **Fig. 2.** Spatial extents of the USMX (dark gray region) and CONUS\_MX (union of dark and
- 651 light gray regions) domains.



**Fig. 3.** Demonstration of gap filling process using LAI over the geographic region bounded by

45-50° N latitude and 120-125° W longitude. (a-c) Maps of 8-day MOD15A2H.006 LAI

aggregated over land cover classes from the MOD12Q1.051 IGBP land cover classification at

657 0.0625° resolution, for days 2000-Oct-07, 2000-Dec-10, and 2001-Feb-10. Gaps are present due 658 to low solar angle, snow, and clouds. (d) Time series of LAI for mixed forest, from the grid cell

to low solar angle, snow, and clouds. (d) Time series of LAI for mixed forest, from the grid cell
denoted by "A" in panels (a-c). (e-f) Climatological mean and standard deviation of LAI for each

- 660 8-day interval over the period 2000-2016. (g) Anomalies of LAI relative to the climatological
- standard deviation, and anomalies after gap filling. (k) Gap-filled time series created by
- 663 recombining gap-filled mean, standard deviation, and anomalies of panels (h-j). (l-n) Gap-filled
- 664 versions of the maps in panels (a-c).





**Fig. 4.** Comparison of MOD-LSP *f*<sub>canopy</sub> from NLCD\_INEGI.2011.2000\_2016 parameters to

668 NLCD canopy cover products aggregated to  $0.0625^{\circ}$  (6 km) resolution. (a) MOD-LSP  $f_{canopy}$ 

from shrubland and grassland classes vs. NLCD-Shrub (Xian et al. 2015); (b) MOD-LSP *f*<sub>canopy</sub>
 from all forest classes vs NLCD-Forest product (Coulston et al. 2012).

671



- **Fig. 5.** Comparison of geographic distributions of broad land cover categories in the L2015,
- 675 MOD\_IGBP, and NLCD\_INEGI.2011 parameter sets. Maps show the total area coverage
- 676 fraction of (a-c) all forest classes; (d-f) all shrub classes; (g-i) all grassland classes; (j-l) all
- 677 agricultural and pastoral classes.





**Fig. 6.** Comparison of climatological average July phenology between L2015 and MOD-LSP

- 680 MOD\_IGBP datasets for the mixed forest class. (a-c) LAI, *f*<sub>canopy</sub>, and albedo from L2015; (d-f)
- 681 LAI, *fcanopy*, and albedo from MOD-LSP MOD\_IGBP.





**Fig. 7.** Comparison of climatological average July phenology between L2015 and MOD-LSP

- 685 MOD\_IGBP datasets for the cropland class. (a-c) LAI, *f*<sub>canopy</sub>, and albedo from L2015; (d-f) LAI,
- *fcanopy*, and albedo from MOD-LSP MOD\_IGBP.
- 687



**Fig. 8.** Comparison of climatological average January and July phenology between L2015 and

- 691 MOD-LSP MOD\_IGBP datasets for the area-weighted average across all land cover classes. (a-
- 692 f) LAI, *f<sub>canopy</sub>*, and albedo from L2015, for (a-c) January and (d-f) July; (g-l) LAI, *f<sub>canopy</sub>*, and
- $\label{eq:general} \mbox{693} \qquad \mbox{albedo from MOD-LSP MOD}\_IGBP, \mbox{ for } (g\mbox{-}i) \mbox{ January and } (j\mbox{-}l) \mbox{ July}.$



695

696 Fig. 9. Comparison of annual hydrologic terms between simulations using the L2015 and MOD-

- 697 LSP MOD\_IGBP datasets over the period 1981-2013 (for *f*<sub>canopy</sub> uniformly held to 1.0). (a-b)
- 698 Mean annual P and T from L2015 meteorological forcings; (c) difference in forest area fraction
- between L2015 and MOD-LSP MOD\_IGBP; (d-f) mean annual LAI from L2015 and
- MOD\_IGBP simulations and their difference; (g-i) mean annual ET from L2015 and
- 701 MOD\_IGBP simulations and their difference; (j-l) mean annual Q from L2015 and MOD\_IGBP
- simulations and their difference. Dashed lines denote boundary between cold/dry climates and
- 703 warm/wet climates.



705

**Fig. 10.** Comparison of annual hydrologic terms between simulations with uniform and spatially-

- varying  $f_{canopy}$ , using the MOD-LSP MOD\_IGBP dataset over the period 1981-2013. (a-c)
- 708 Uniform  $f_{canopy}$ , mean annual spatially-varying  $f_{canopy}$ , and their difference; (d-f) mean annual ET
- from uniform and spatially-varying  $f_{canopy}$  simulations and their difference; (h-j) mean annual Q from uniform and spatially-varying  $f_{canopy}$  simulations and their difference; (g) and (k) total
- from uniform and spatially-varying *f<sub>canopy</sub>* simulations and their difference; (g) and (k) total
   difference in (g) ET and (k) Q between MOD-IGBP spatially-varying *f<sub>canopy</sub>* and L2015 uniform
- $f_{canopy}$  simulations. Dashed lines denote boundary between cold/dry climates and warm/wet
- 713 climates.
- 714



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**Fig. 11.** Changes in the geographic distributions of broad land cover classes between 1992 and

2011 in the NLCD\_INEGI parameter sets. Panel (a) Distribution of all forest classes in 2011; (b)
change in distribution of forest between 1992 and 2011; (c-f) magnification of boxes 1-4 in panel

(b). Panels (g-l) same as (a-f) for combined areas of shrubland and grassland classes. Panels (m-

r) same as (a-f) for combined areas of agricultural and pastoral classes. Panels (s-x) same as (a-f)

- 721 for combined areas of urban classes.
- 722



- Fig. 12. Changes in the phenology and annual hydrologic fluxes between 2001 and 2011 in the
- NLCD\_INEGI.2001.2001\_2001 and NLCD\_INEGI.2011.2011\_2011 parameter sets. Panel (a)
- mean annual LAI in 2011; (b) change in mean annual LAI between 2001 and 2011; (c-f)
- magnification of boxes 1-4 in panel (b). Panels (g-l) same as (a-f) for mean annual *f*<sub>canopy</sub>. Panels
- 728 (m-r) same as (a-f) for mean annual ET. Panels (s-x) same as (a-f) for mean annual Q.
- 729
- 730