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### Supplement of

### Creative computing with Landlab: an open-source toolkit for building, coupling, and exploring two-dimensional numerical models of Earth-surface dynamics

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### This document contains:

**Table S1:** A list of all Landlab standard names in Landlab v.1.0.

Scripts S2-S6: Scripts to run the models described in section 5 of the main text. Note

S5 also contains the text of a separate input file after the code block.

Table S1. Field names used by Landlab version 1.0.

Table S1. Field names used by La		
Field name	Used by	Provided by
channel_bed_shear_stress		SedDepEroder
channel_chi_index		ChiFinder
channel_depth		SedDepEroder
channel_discharge		SedDepEroder
channel_steepness_index		SteepnessFinder
channel_width		SedDepEroder
channel_sediment_relative_flux		SedDepEroder
channel_sediment_volumetric_flux		SedDepEroder
channel_sediment_volumetric_transport_capacity	D : E: 1 4 1D :	SedDepEroder
depression_depth	DepressionFinderAndRouter	
depression_outlet_node	DepressionFinderAndRouter	ElDt
drainage_area	ChiFinder	FlowRouter
	FastscapeEroder	
	SedDepEroder	
	SteepnessFinder StreamPowerEroder	
G lil- ti l	ChiFinder	FlowRouter
flow_link_to_receiver_node		FlowRouter
	FastscapeEroder	
	SedDepEroder SteepnessFinder	
	StreamPowerEroder	
flow_receiver_node	ChiFinder	FlowRouter
now_receiver_node	FastscapeEroder	riowkoutei
	SedDepEroder	
	SteepnessFinder	
	StreamPowerEroder	
flow_sink_flag	Stream ower Er oder	FlowRouter
flow_upstream_node_order	ChiFinder	FlowRouter
	FastscapeEroder	
	SedDepEroder	
	SteepnessFinder	
	StreamPowerEroder	
lithosphere_overlying_pressure_increment	Flexure	
lithosphere_surface_elevation_increment		Flexure
•		gFlex
plant_age		VegCA
plant_live_index		VegCA
radiation_incoming_shortwave_flux		PotentialEvapotranspiration
		Radiation
radiation_net_flux		PotentialEvapotranspiration
radiation_net_longwave_flux		PotentialEvapotranspiration
radiation_net_shortwave_flux		PotentialEvapotranspiration
		Radiation
radiationratio_to_flat_surface	PotentialEvapotranspiration	Radiation
rainfall_daily_depth	SoilMoisture	
sediment_fill_depth		SinkFiller
soil_moistureinitial_saturation_fraction	SoilMoisture	
soil_moisture_root_zone_leakage		SoilMoisture
soil_moisture_saturation_fraction		SoilMoisture
soil_water_infiltrationdepth	SoilInfiltrationGreenAmpt	SoilInfiltrationGreenAmpt
curtage evanetranguiration	Vegetation	SoilMoisture
surface_evapotranspiration	37	
surface_potential_evapotranspiration_30day_mean	Vegetation	D. C. ID.
	SoilMoisture	PotentialEvapotranspiration
surface_potential_evapotranspiration_30day_mean	_	PotentialEvapotranspiration SoilMoisture

(continues)

(continued)

Field name	Used by	Provided by
surface_loadstress	gFlex	
surface_waterdepth	KinematicWaveRengers	KinematicWaveRengers
	OverlandFlow	OverlandFlow
	OverlandFlowBates	OverlandFlowBates
	SoilInfiltrationGreenAmpt	SoilInfiltrationGreenAmpt
surface_water_discharge	DetachmentLtdErosion	FlowRouter KinematicWaveRengers
		OverlandFlow
		OverlandFlowBates
surface_water_velocity	ol ini	KinematicWaveRengers
topographic_elevation	ChiFinder	DetachmentLtdErosion
	DepressionFinderAndRouter	FastscapeEroder
	DetachmentLtdErosion	gFlex
	FastscapeEroder	LinearDiffuser
	FlowRouter	PerronNLDiffuse
	KinematicWaveRengers	SedDepEroder
	LinearDiffuser	SinkFiller
	OverlandFlow	StreamPowerEroder
	OverlandFlowBates	
	PerronNLDiffuse	
	Radiation	
	SedDepEroder	
	SinkFiller	
	SteepnessFinder	
	StreamPowerEroder	
topographic_gradient		LinearDiffuser
topographic_slope	DetachmentLtdErosion	
topographic_steepest_slope	ChiFinder	FlowRouter
	SedDepEroder	
	SteepnessFinder	
	StreamPowerEroder	
hillslope_sediment_unit_volume_flux		LinearDiffuser
vegetation_cover_fraction	SoilMoisture	
vegetation_cumulative_water_stress	VegCA	Vegetation
vegetation_dead_biomass		Vegetation
vegetation_dead_leaf_area_index		Vegetation
vegetation_live_biomass		Vegetation
vegetation_live_leaf_area_index	SoilMoisture	Vegetation
vegetation_plant_functional_type	SoilMoisture VegCA	
	Vegetation	
vegetation_water_stress	Vegetation	SoilMoisture
water_unit_flux_in	FlowRouter	
water_surfacegradient		OverlandFlow
		OverlandFlowBates

# Script S2. Code to examine run times of simple stream power models on two Landlab grid types (Fig. 11).

```
1
     import numpy as np
     from matplotlib.pyplot import figure, show, plot, xlabel, ylabel, ylim
2
     from landlab import RasterModelGrid, HexModelGrid
3
     from landlab.components import StreamPowerEroder, FlowRouter, \
4
5
         PrecipitationDistribution
    from landlab import imshow grid
6
7
     from time import time
8
9
     uplift rate = 0.001
10
     side list = [5, 10, 15, 20, 30, 40, 50, 75, 100, 150]
11
    total time listr = []
12
     total time listh = []
13
     loop time listr = []
14
     loop time listh = []
15
     num repeats = 5
16
17
     for side in side_list:
18
         print(side)
19
         for i in range(2):
20
             temp tottime = []
21
             temp looptime = []
22
             for j in range(num repeats):
23
                 time 0 = time()
24
                 if i == 0:
25
                     mg = RasterModelGrid((side, side), 100.)
26
                 else:
27
                     mg = HexModelGrid(side, side, 100., shape='rect')
28
29
                 mg_noise = np.random.rand(mg.number_of_nodes)/1000.
30
31
                 zr = mg.add zeros('node', 'topographic elevation')
32
                 zr += mg noise
33
                 Qr = mg.add empty('node', 'surface water discharge')
34
35
                 runoff rater = mg.add ones('node', 'water unit flux in')
36
37
                 frr = FlowRouter(mg)
                 # ^water__unit_flux_in gets automatically ingested
38
39
                 spr = StreamPowerEroder(mg, K sp=1.e-5, threshold sp=1.e-6,
40
                                          use Q='surface water discharge')
41
                 precip = PrecipitationDistribution(
42
                     mean storm depth=5000., mean storm duration=100.,
43
                     mean interstorm duration=900., total t=3.e6)
44
45
                 time in = time()
46
                 for (dt, runoff) in \
47
                         precip.yield storm interstorm duration intensity():
48
                     zr[mq.core nodes] += uplift rate*dt
49
                     if runoff > 0.:
50
                         runoff rater.fill(runoff)
51
                         frr.run one step()
52
                         spr.run_one_step(dt)
```

```
53
                          # raincount += 1
54
                          # print(raincount, dt, runoff)
55
                 time out = time()
56
                 temp tottime.append(time out-time 0)
57
                 temp looptime.append(time out-time in)
58
             tottime = np.mean(temp tottime)
59
             looptime = np.mean(temp looptime)
60
             if i == 0:
                 loop time listr.append(looptime)
61
62
                 total time listr.append(tottime)
63
             else:
64
                 loop time listh.append(looptime)
65
                 total time listh.append(tottime)
66
     np.savetxt('side_list', np.array(side_list))
67
68
     np.savetxt('loop_time_listr', np.array(loop_time_listr))
69
     np.savetxt('total_time_listr', np.array(total_time_listr))
70
     np.savetxt('loop time listh', np.array(loop time listh))
71
     np.savetxt('total time listh', np.array(total time listh))
72
     side list = np.loadtxt('side_list')
73
74
     loop time listr = np.loadtxt('loop time listr')
75
     total_time_listr = np.loadtxt('total_time_listr')
76
     loop_time_listh = np.loadtxt('loop_time_listh')
77
     total time listh = np.loadtxt('total time listh')
78
     s = 12.
79
     figure('runtimes')
     plot(np.array(side_list)**2, loop_time_listr, 'b+', markersize=s)
80
81
     plot(np.array(side list)**2, total time listr, 'D',
82
          markerfacecolor='none', markeredgecolor='b', markersize=s)
83
     plot(np.array(side list)**2, loop time listh, 'rx', markersize=s)
84
     plot(np.array(side list)**2, total time listh, 's',
85
          markerfacecolor='none', markeredgecolor='r', markersize=s)
86
     xlabel('Number of nodes')
87
     ylabel('Time (s)')
88
89
     figure('overhead')
90
     overhead r = np.array(total time listr) - np.array(loop time listr)
91
     overhead_h = np.array(total_time_listh) - np.array(loop_time_listh)
     plot(np.array(side_list)**2, overhead_r, 'bo', markersize=s)
92
93
     plot(np.array(side_list)**2, overhead_h, 'r^', markersize=s)
94
     ylim([0, 2.1])
     xlabel('Number of nodes')
95
96
     ylabel('Time (s)')
97
98
     show()
```

# Script S3. Code to run a simple stream power model in Landlab, incorporating both storms and a threshold, on two different grid types (Fig. 12).

```
1
     import numpy as np
     from matplotlib.pyplot import figure, show, loglog, xlim, ylim, xlabel, ylabel
2
     from landlab import RasterModelGrid, HexModelGrid
3
     from landlab.components import StreamPowerEroder, FlowRouter, \
4
5
         PrecipitationDistribution
     from landlab import imshow grid
6
7
     from copy import deepcopy
8
9
     # This script is to make fig 11
10
     side = 100
11
     uplift rate = 0.001
     gridlist = []
12
13
14
     for i in range(2):
15
         if i == 0:
16
             mg = RasterModelGrid((side, side), 100.)
17
         else:
             mg = HexModelGrid(side, side, 100., shape='rect')
18
19
20
         # add initial noise to produce convergent flow from the initial conditions
         np.random.seed(0) # so our figures are reproducible
21
22
         mg noise = np.random.rand(mg.number of nodes)/1000.
23
         # set up the input fields
24
25
         zr = mg.add_zeros('node', 'topographic__elevation')
         zr += mg noise
26
27
         Qr = mg.add empty('node', 'surface water discharge')
         runoff_rater = mg.add_ones('node', 'water__unit_flux_in')
28
29
         # Landlab sets fixed elevation boundary conditions by default. This is
31
         # what we want, so we will not modify these here.
32
33
         # instantiate the components:
34
         frr = FlowRouter(mg) # water__unit_flux_in gets automatically ingested
35
         spr = StreamPowerEroder(
             mg, K_sp=1.e-5, m_sp=0.5, n_sp=1., threshold_sp=1.e-6,
36
             use_Q='surface_water__discharge')
37
         # the `use Q` flag tells the StreamPowerEroder to use discharge defined in
38
39
         # the field 'surface_water__discharge' as the first term in the stream
40
         # power law, not the drainage area, as is sometimes also seen.
         precip = PrecipitationDistribution(
41
42
             mean storm depth=5000., mean storm duration=100.,
43
             mean_interstorm_duration=900., total_t=3.e6)
44
45
         raincount = 0 # this flag lets us see how many rain events have occured
46
         for (dt, runoff) in precip.yield storm interstorm duration intensity():
             zr[mg.core nodes] += uplift rate*dt
47
48
             if runoff > 0.:
49
                 runoff_rater.fill(runoff)
50
                 frr.run_one_step()
51
                 spr.run one step(dt)
52
                 raincount += 1
```

```
53
                  print(raincount, dt, runoff)
 54
              # this loop will terminate automatically, thanks to the generator
              # method we're calling from the `precip` class object.
 55
 56
 57
          # Do some plotting. First the topography:
 58
          figure('topo ' + str(i))
 59
          imshow grid(mg, zr, grid units=('m', 'm'), var name='Elevation (m)')
 60
          # then some slope-area plots, for checking:
 61
 62
          figure('S-A all')
 63
          loglog(mg.at_node['drainage_area'],
                 mg.at node['topographic steepest slope'], 'x')
 64
 65
          figure('S-A interior only ' + str(i))
          edge = np.unique(mg.neighbors_at_node[mg.boundary_nodes, :])
 66
 67
          not_edge = np.in1d(mg.nodes.flatten(), edge, assume_unique=True,
 68
                              invert=True)
 69
          loglog(mg.at node['drainage area'][not edge],
 70
                 mq.at node['topographic steepest slope'][not edge], 'x')
 71
          xlim([1.e3, 1.e8])
 72
          xlabel('Topographic slope')
          ylabel('Drainage area (m^2)')
 73
 74
 75
          # save the data to a list so we can inspect both grid types at our leisure
 76
          # if this script is run from an interactive Python environment like
 77
          # iPython:
 78
          gridlist.append(deepcopy(mg))
 79
 80
      # show all the plots we have:
 81
      show()
 82
 83
      # just to produce the figures:
 84
      for mg in gridlist:
 85
 86
          figure('topo ' + str(i))
 87
          imshow_grid(mg, 'topographic__elevation', grid_units=('m', 'm'),
 88
                      var name='Elevation (m)')
 89
          figure('S-A ' + str(i))
 90
          edge = np.unique(mg.neighbors_at_node[mg.boundary_nodes, :])
 91
          not edge = np.inld(mg.nodes.flatten(), edge, assume unique=True,
 92
                              invert=True)
 93
          loglog(mg.at_node['drainage_area'][not_edge],
 94
                 mg.at node['topographic steepest slope'][not edge], 'x')
 95
          xlim([1.e3, 1.e7])
 96
          ylabel('Topographic slope')
 97
          xlabel('Drainage area (m^2)')
 98
          figure('S-A')
          loglog(mg.at node['drainage area'][not edge],
 99
100
                 mg.at_node['topographic__steepest_slope'][not_edge], 'x')
101
          xlim([1.e3, 1.e7])
102
          ylabel('Topographic slope')
103
          xlabel('Drainage area (m^2)')
104
          i += 1
105
      show()
106
```

# Script S4. Code to run a coupled stream power-hillslope diffusion model in Landlab on two different grid types (Fig. 12).

```
import numpy as np
1
     from matplotlib.pyplot import figure, show, loglog, xlim, ylim, xlabel, ylabel
2
     from landlab import RasterModelGrid, HexModelGrid
3
     from landlab.components import StreamPowerEroder, FlowRouter, \
4
         PrecipitationDistribution, LinearDiffuser, DepressionFinderAndRouter
5
6
     from landlab import imshow grid
7
     from copy import deepcopy
8
9
     side = 100
10
     uplift rate = 0.001
11
     gridlist = []
12
13
     for i in range(2):
14
         if i == 0:
15
             mg = RasterModelGrid((side, side), 100.)
16
         else:
             mg = HexModelGrid(side, side, 100., shape='rect')
17
18
19
         # add initial noise to produce convergent flow from the initial conditions
20
         np.random.seed(0) # so our figures are reproducible
         mg noise = np.random.rand(mg.number of nodes)/1000.
21
22
23
         # set up the input fields
         zr = mg.add_zeros('node', 'topographic__elevation')
24
25
         zr += mg noise
         Qr = mg.add_empty('node', 'surface_water__discharge')
26
27
         runoff rater = mg.add ones('node', 'water unit flux in')
28
29
         # Landlab sets fixed elevation boundary conditions by default. This is
         # what we want, so we will not modify these here.
31
32
         # instantiate the components:
33
         frr = FlowRouter(mg) # water unit flux in gets automatically ingested
34
         spr = StreamPowerEroder(
35
             mg, K sp=1.e-5, m sp=0.5, n sp=1., threshold sp=1.e-6,
             use Q='surface water discharge')
36
         # the `use Q` flag tells the StreamPowerEroder to use discharge defined in
37
         # the field 'surface_water__discharge' as the first term in the stream
38
39
         # power law, not the drainage area, as is sometimes also seen.
40
         dfn = LinearDiffuser(mg, linear_diffusivity=0.05)
         lake = DepressionFinderAndRouter(mg)
41
42
         precip = PrecipitationDistribution(
43
             mean_storm_depth=5000., mean_storm_duration=100.,
             mean interstorm duration=900., total t=3.e6)
44
45
46
         raincount = 0 # this flag lets us see how many rain events have occured
47
         for (dt, runoff) in precip.yield storm interstorm duration intensity():
48
             zr[mq.core nodes] += uplift rate*dt
49
             dfn.run_one_step(dt) # hillslopes always diffusive, even when dry
50
             if runoff > 0.:
51
                 runoff rater.fill(runoff)
52
                 frr.run_one_step()
```

```
53
                  lake.map depressions()
 54
                  spr.run_one_step(dt, flooded_nodes=lake.lake_at_node)
 55
                  raincount += 1
 56
                  print(raincount, dt, runoff)
 57
              # this loop will terminate automatically, thanks to the generator
 58
              # method we're calling from the `precip` class object.
 59
 60
          # Do some plotting. First the topography:
          figure('topo ' + str(i))
 61
          imshow grid(mg, zr, grid units=('m', 'm'), var name='Elevation (m)')
 62
 63
          # then some slope-area plots, for checking:
 64
 65
          figure('S-A all')
          loglog(mg.at_node['drainage_area'],
 66
                 mg.at_node['topographic__steepest_slope'], 'x')
 67
          figure('S-A_interior_only ' + str(i))
 68
 69
          edge = np.unique(mg.neighbors at node[mg.boundary nodes, :])
 70
          not edge = np.inld(mg.nodes.flatten(), edge, assume unique=True,
 71
                              invert=True)
 72
          loglog(mg.at node['drainage area'][not edge],
 73
                 mg.at node['topographic steepest slope'][not edge], 'x')
 74
          xlim([1.e3, 1.e7])
 75
          xlabel('Topographic slope')
 76
          ylabel('Drainage area (m^2)')
 77
 78
          # save the data to a list so we can inspect both grid types at our leisure
 79
          # if this script is run from an interactive Python environment like
 80
          # iPython:
 81
          gridlist.append(deepcopy(mg))
 82
 83
      # show all the plots we have:
 84
      show()
 85
 86
      # just to produce the figures:
 87
      i = 0
 88
      for mg in gridlist:
 89
          figure('topo ' + str(i))
 90
          imshow grid(mg, 'topographic elevation', grid units=('m', 'm'),
 91
                      var name='Elevation (m)')
 92
          figure('S-A ' + str(i))
          edge = np.unique(mg.neighbors_at_node[mg.boundary_nodes, :])
 93
 94
          not edge = np.inld(mg.nodes.flatten(), edge, assume unique=True,
 95
                              invert=True)
 96
          loglog(mg.at node['drainage area'][not edge],
                 mg.at node['topographic steepest slope'][not edge], 'x')
 97
 98
          xlim([1.e3, 1.e7])
          ylabel('Topographic slope')
 99
100
          xlabel('Drainage area (m^2)')
101
          figure('S-A')
102
          loglog(mg.at node['drainage area'][not edge],
103
                 mg.at node['topographic steepest slope'][not edge], 'x')
104
          xlim([1.e3, 1.e7])
105
          ylabel('Topographic slope')
106
          xlabel('Drainage area (m^2)')
          i += 1
107
108
      show()
```

#### Script S5. Code to run an ecohydrology model in Landlab.

```
# Authors: Sai Nudurupati & Erkan Istanbulluoglu, 21May15
1
     # Edited: 15Jul16 - to conform to Landlab version 1.
2
     # A companion interactive tutorial an be found at: landlab/tutorials/
     # ...ecohydrology/cellular automaton vegetation flat surface.ipynb
5
6
     import os
7
     import time
     import numpy as np
8
     import matplotlib as mpl
9
10
     import matplotlib.pyplot as plt
11
     from landlab import load params, RasterModelGrid
12
13
     from landlab.plot import imshow grid
     from landlab.components import (PrecipitationDistribution, Radiation,
14
15
                                      PotentialEvapotranspiration, SoilMoisture,
16
                                      Vegetation, VegCA)
17
18
     GRASS = 0
19
    SHRUB = 1
20
    TREE = 2
21
    BARE = 3
22
     SHRUBSEEDLING = 4
23
     TREESEEDLING = 5
24
25
26
     def compose veg grid(grid, percent bare=0.4, percent grass=0.2,
27
                          percent shrub=0.2, percent tree=0.2):
28
         """Compose spatially distribute PFT."""
29
         no cells = grid.number of cells
30
         shrub point = int(percent bare * no cells)
31
         tree point = int((percent bare + percent shrub) * no cells)
32
         grass point = int((1 - percent grass) * no cells)
33
         veg grid = np.full(grid.number of cells, BARE, dtype=int)
35
         veg grid[shrub point:tree point] = SHRUB
36
         veg grid[tree point:grass point] = TREE
37
         veg grid[grass point:] = GRASS
38
39
         np.random.shuffle(veg grid)
40
         return veg_grid
41
42
43
     def initialize(data, grid, grid1):
44
         """Initialize random plant type field.
45
         Plant types are defined as the following:
46
47
48
         * GRASS = 0
         * SHRUB = 1
49
50
         * TREE = 2
         * BARE = 3
51
           SHRUBSEEDLING = 4
52
53
         * TREESEEDLING = 5
         ....
54
```

```
55
          grid1.at cell['vegetation plant functional type'] = compose veg grid(
 56
              grid1, percent_bare=data['percent_bare_initial'],
              percent_grass=data['percent_grass_initial'],
 57
 58
              percent_shrub=data['percent_shrub_initial'],
 59
              percent tree=data['percent tree initial'])
 60
 61
          # Assign plant type for representative ecohydrologic simulations
 62
          grid.at cell['vegetation plant functional type'] = np.arange(6)
          grid1.at node['topographic elevation'] = np.full(grid1.number of nodes,
 63
 64
                                                             1700.)
 65
          grid.at_node['topographic__elevation'] = np.full(grid.number_of_nodes,
 66
                                                            1700.)
 67
          precip dry = PrecipitationDistribution(
              mean_storm_duration=data['mean_storm_dry'],
 68
              mean_interstorm_duration=data['mean_interstorm_dry'],
 69
 70
              mean_storm_depth=data['mean_storm_depth_dry'])
 71
          precip wet = PrecipitationDistribution(
 72
              mean storm duration=data['mean storm wet'],
 73
              mean interstorm duration=data['mean interstorm wet'],
 74
              mean storm depth=data['mean storm depth wet'])
 75
 76
          radiation = Radiation(grid)
 77
          pet tree = PotentialEvapotranspiration(grid, method=data['PET method'],
 78
                                                  MeanTmaxF=data['MeanTmaxF_tree'],
 79
                                                  delta d=data['DeltaD'])
 80
          pet shrub = PotentialEvapotranspiration(grid, method=data['PET method'],
                                                   MeanTmaxF=data['MeanTmaxF shrub'],
 81
 82
                                                   delta d=data['DeltaD'])
 83
          pet grass = PotentialEvapotranspiration(grid, method=data['PET method'],
 84
                                                   MeanTmaxF=data['MeanTmaxF grass'],
 85
                                                   delta d=data['DeltaD'])
 86
          soil moisture = SoilMoisture(grid, **data) # Soil Moisture object
          vegetation = Vegetation(grid, **data) # Vegetation object
 87
          vegca = VegCA(grid1, **data) # Cellular automaton object
 88
 89
 90
          # Initializing inputs for Soil Moisture object
 91
          grid.at cell['vegetation live leaf area index'] = (
 92
              1.6 * np.ones(grid.number of cells))
          grid.at cell['soil moisture initial saturation fraction'] = (
 93
 94
              0.59 * np.ones(grid.number of cells))
 95
 96
          return (precip dry, precip wet, radiation, pet tree, pet shrub,
 97
                  pet grass, soil moisture, vegetation, vegca)
 98
99
      def empty arrays(n, grid, grid1):
100
          precip = np.empty(n) # Record precipitation
101
102
          inter storm dt = np.empty(n) # Record inter storm duration
103
          storm dt = np.empty(n) # Record storm duration
104
          time elapsed = np.empty(n) # To record time elapsed from start of simulation
105
106
          # Cumulative Water Stress
107
          veg_type = np.empty([n / 55, grid1.number_of_cells], dtype=int)
108
          daily_pet = np.zeros([365, grid.number_of_cells])
109
          rad factor = np.empty([365, grid.number of cells])
110
          EP30 = np.empty([365, grid.number of cells])
```

```
111
112
          # 30 day average PET to determine season
          pet_threshold = 0 # Initializing pet_threshold to ETThresholddown
113
          return (precip, inter storm dt, storm dt, time elapsed, veg type,
114
                   daily pet, rad factor, EP30, pet threshold)
115
116
117
118
      def create pet lookup(radiation, pet tree, pet shrub, pet grass, daily pet,
                             rad factor, EP30, grid):
119
120
          for i in range(0, 365):
121
              pet tree.update(float(i) / 365.25)
               pet shrub.update(float(i) / 365.25)
122
               pet grass.update(float(i) / 365.25)
123
               daily_pet[i] = [pet_grass._PET_value, pet_shrub._PET_value,
124
                          pet_tree._PET_value, 0., pet_shrub._PET_value,
125
126
                          pet_tree._PET_value]
127
              radiation.update(float(i) / 365.25)
              rad factor[i] = grid.at cell['radiation ratio to flat surface']
128
129
              if i < 30:
130
                   if i == 0:
131
132
                       EP30[0] = daily pet[0]
133
                   else:
                       EP30[i] = np.mean(daily_pet[:i], axis=0)
134
               else:
135
136
                  EP30[i] = np.mean(daily pet[i - 30:i], axis=0)
137
138
139
      def save(sim, inter storm dt, storm dt, precip, veg type, yrs,
140
               walltime, time elapsed):
141
          np.save(sim + ' Tb', inter storm dt)
          np.save(sim + '_Tr', storm_dt)
142
          np.save(sim + '_P', precip)
np.save(sim + '_VegType', veg_type)
143
144
          np.save(sim + '_Years', yrs)
145
          np.save(sim + '_Time_Consumed_minutes', walltime)
146
147
          np.save(sim + ' CurrentTime', time elapsed)
148
149
150
      def plot(sim, grid, veg_type, yrs, yr_step=10):
151
          pic = 0
152
          years = range(0, yrs)
153
          cmap = mpl.colors.ListedColormap(
154
              ['green', 'red', 'black', 'white', 'red', 'black'])
          bounds = [-0.5, 0.5, 1.5, 2.5, 3.5, 4.5, 5.5]
155
156
          norm = mpl.colors.BoundaryNorm(bounds, cmap.N)
157
          print 'Plotting cellular field of Plant Functional Type'
158
          print 'Green - Grass; Red - Shrubs; Black - Trees; White - Bare'
159
160
          # Plot images to make gif.
161
          for year in range(0, yrs, yr_step):
               filename = 'year ' + "%05d" % year
162
163
              pic += 1
164
              plt.figure(pic, figsize=(10, 8))
165
               imshow grid(grid, veg type[year], values at='cell', cmap=cmap,
                           grid_units=('m', 'm'), norm=norm, limits=[0, 5],
166
```

```
167
                          allow colorbar=False)
              plt.title(filename, weight='bold', fontsize=22)
168
              plt.xlabel('X (m)', weight='bold', fontsize=18)
169
170
              plt.ylabel('Y (m)', weight='bold', fontsize=18)
              plt.xticks(fontsize=14, weight='bold')
171
172
              plt.yticks(fontsize=14, weight='bold')
173
              plt.savefig(sim + ' ' + filename)
174
175
          grass cov = np.empty(yrs)
176
          shrub cov = np.empty(yrs)
177
          tree_cov = np.empty(yrs)
178
          grid size = float(veg type.shape[1])
179
180
          for x in range(0, yrs):
              grass_cov[x] = (veg_type[x][veg_type[x] == GRASS].size /
181
182
                              grid size) * 100
              shrub_cov[x] = ((veg_type[x][veg_type[x] == SHRUB].size / grid_size) *
183
                               100 + (veg_type[x][veg_type[x] == SHRUBSEEDLING].size /
184
185
                              grid size) * 100)
              tree cov[x] = ((veg type[x][veg type[x] == TREE].size / grid size) *
186
                              100 + (veg_type[x][veg_type[x] == TREESEEDLING].size /
187
188
                              grid size) * 100)
189
190
          pic += 1
          plt.figure(pic, figsize=(10, 8))
191
192
          plt.plot(years, grass cov, '-g', label='Grass', linewidth=4)
193
          plt.hold(True)
          plt.plot(years, shrub cov, '-r', label='Shrub', linewidth=4)
194
195
          plt.hold(True)
196
          plt.plot(years, tree cov, '-k', label='Tree', linewidth=4)
197
          plt.ylabel('% Area Covered by Plant Type', weight='bold', fontsize=18)
198
          plt.xlabel('Time in years', weight='bold', fontsize=18)
199
          plt.xticks(fontsize=12, weight='bold')
200
          plt.yticks(fontsize=12, weight='bold')
          plt.legend(loc=0, prop={'size': 16, 'weight': 'bold'})
201
202
          plt.savefig(sim + '_percent_cover')
203
204
205
      # Now a script to drive the model:
206
      grid1 = RasterModelGrid((100, 100), spacing=(5., 5.))
207
      grid = RasterModelGrid((5, 4), spacing=(5., 5.))
208
209
210
      # Create dictionary that holds the inputs
      data = load params('inputs vegetation ca.yaml')
211
212
213
      (precip dry, precip wet, radiation, pet tree, pet shrub,
214
       pet grass, soil moisture, vegetation, vegca) = initialize(data, grid, grid1)
215
216
      n years = 2000 # Approx number of years for model to run
217
218
      # Calculate approximate number of storms per year
219
      fraction_wet = (data['doy__end_of_monsoon'] -
220
                      data['doy__start_of_monsoon']) / 365.
221
      fraction dry = 1 - fraction wet
222
      no_of_storms_wet = 8760 * fraction_wet / (data['mean_interstorm_wet'] +
```

```
223
                                                 data['mean storm wet'])
224
      no_of_storms_dry = 8760 * fraction_dry / (data['mean_interstorm_dry'] +
225
                                                 data['mean_storm_dry'])
226
      n = int(n years * (no of storms wet + no of storms dry))
227
228
      (precip, inter storm dt, storm dt, time elapsed, veg type, daily pet,
      rad factor, EP30, pet threshold) = empty_arrays(n, grid, grid1)
229
230
231
      create pet lookup(radiation, pet tree, pet shrub, pet grass, daily pet,
232
                        rad factor, EP30, grid)
233
234
      # Represent current time in years
235
      current time = 0 # Start from first day of Jan
236
237
      # Keep track of run time for simulation - optional
238
      wallclock start = time.clock() # Recording time taken for simulation
239
240
      # declaring few variables that will be used in the storm loop
241
      time check = 0. # Buffer to store current time at previous storm
242
      yrs = 0 # Keep track of number of years passed
243
      water stress = 0. # Buffer for Water Stress
244
      Tg = 270 # Growing season in days
245
246
      # Run storm Loop
247
      for i in range(n):
248
          # Update objects
249
250
          # Calculate Day of Year (DOY)
251
          julian = np.int(np.floor((current time - np.floor(current time)) * 365.))
252
253
          # Generate seasonal storms
254
          # Wet Season - Jul to Sep - NA Monsoon
          if data['doy start of monsoon'] <= julian <= data['doy end of monsoon']:
255
256
              precip wet.update()
257
              precip[i] = precip_wet.storm_depth
258
              storm dt[i] = precip wet.storm duration
259
              inter storm dt[i] = precip wet.interstorm duration
260
          else: # for Dry season
261
              precip dry.update()
262
              precip[i] = precip_dry.storm_depth
263
              storm_dt[i] = precip_dry.storm_duration
264
              inter storm dt[i] = precip dry.interstorm duration
265
          # Spatially distribute PET and its 30-day-mean (analogous to degree day)
266
267
          grid.at cell[
              'surface potential evapotranspiration rate' | = daily pet[julian]
268
269
          grid.at cell[
270
              'surface potential evapotranspiration 30day mean' | = EP30[julian]
271
272
          # Assign spatial rainfall data
          grid.at_cell[
273
274
              'rainfall daily depth'] = np.full(grid.number of cells, precip[i])
275
276
          # Update soil moisture component
277
          current time = soil moisture.update(current time, Tr=storm dt[i],
278
                                              Tb=inter storm dt[i])
```

```
279
          # Decide whether its growing season or not
280
          if julian != 364:
281
282
              if EP30[julian + 1, 0] > EP30[julian, 0]:
283
                  pet threshold = 1
284
                  # 1 corresponds to ETThresholdup (begin growing season)
285
              else:
286
                  pet threshold = 0
                  # 0 corresponds to ETThresholddown (end growing season)
287
288
289
          # Update vegetation component
290
          vegetation.update(PETThreshold switch=pet threshold, Tb=inter storm dt[i],
291
                            Tr=storm dt[i])
292
293
          # Update yearly cumulative water stress data
294
          water_stress += (grid.at_cell['vegetation__water_stress'] *
295
                           inter storm dt[i] / 24.)
296
297
          # Record time (optional)
298
          time elapsed[i] = current time
299
300
          # Update spatial PFTs with Cellular Automata rules
301
          if (current_time - time_check) >= 1.:
302
              if yrs % 100 == 0:
303
                  print 'Elapsed time = ', yrs, ' years'
304
              veg type[yrs] = grid1.at cell['vegetation plant functional type']
305
              WS_ = np.choose(veg_type[yrs], water_stress)
306
              grid1.at_cell['vegetation__cumulative_water_stress'] = WS_ / Tg
307
              vegca.update()
308
              time check = current time
309
              water stress = 0
310
              yrs += 1
311
312
      veg type[yrs] = grid1.at cell['vegetation plant functional type']
313
314
      wallclock stop = time.clock()
315
      walltime = (wallclock stop - wallclock start) / 60. # in minutes
316
      print 'Time consumed = ', walltime, ' minutes'
317
318
      # Saving
319
      try:
320
          os.mkdir('output')
321
      except OSError:
322
          pass
323
      finally:
          os.chdir('output')
324
325
326
      save('veg', inter_storm_dt, storm_dt, precip, veg_type, yrs,
327
           walltime, time elapsed)
328
329
      plot('veg', grid1, veg type, yrs, yr step=100)
```

This code makes use of the following input text file, named "inputs\_vegetation\_ca.yaml":

```
### All inputs for Vegetation Cellular Automaton Model built on The Landlab
### can be given here.
### 14Feb2015 - Sai Nudurupati & Erkan Istanbulluoglu
### 15Jul2016 - Updated to comply with Landlab Version 1 naming conventions.
### Vegetation Cellular Automaton Model Input File:
n short: 6600 # Number of storms for short simulation that plots hydrologic
parameters
n long DEM: 1320 # Number of storms for long simulation that operates on single
grid for sloped surface
n long flat: 660000 # Number of storms for long simulation that operates on two
grids - flat surface
## Initial Plant Functional Types (PFT) distribution
percent_bare_initial: 0.7 # Initial percentage of cells occupied by bare soil
percent_grass_initial: 0.1 # Initial percentage of cells occupied by grass
percent shrub initial: 0.1 # Initial percentage of cells occupied by shrubs
percent tree initial: 0.1 # Initial percentage of cells occupied by trees
## Precipitation:
# Dry Season
mean storm dry: 2.016 # Mean storm duration (hours)
mean_interstorm_dry: 159.36 # Mean interstorm duration (hours)
mean_storm_depth_dry: 3.07 # Mean storm depth (mm)
# Wet Season
mean_storm_wet: 1.896 # Mean storm duration (hours)
mean interstorm wet: 84.24 # Mean interstorm duration (hours)
mean storm depth wet: 4.79 # Mean storm depth (mm)
doy start of monsoon: 182 # Day of the year when the monsoon starts
doy__end_of_monsoon: 273 # Day of the year when the monsoon ends
## PotentialEvapotranspiration:
# Cosine Method
PET method: Cosine
LT: 0 # Lag between peak TmaxF estimated by cosine method and solar forcing
DeltaD: 7. # Calibrated difference between
ND: 365. # Number of days in the year (days)
MeanTmaxF grass: 5.15 # Mean annual rate of TmaxF (mm/d)
MeanTmaxF shrub: 3.77 # Mean annual rate of TmaxF (mm/d)
MeanTmaxF tree: 4.96 # Mean annual rate of TmaxF (mm/d)
# TmaxF - Estimated maximum evapotranspiration as a function of DOY
# using Penman Monteith method for historical weather
## Soil Moisture:
runon: 0. # Runon from higher elevations (mm)
f bare: 0.7 # Fraction to partition PET for bare soil (None)
```

```
VEGTYPE grass: 0 # Integer value to infer Vegetation Type
intercept cap grass: 1. # Full canopy interception capacity (mm)
zr grass: 0.3 # Root depth (m)
I B grass: 20. # Infiltration capacity of bare soil (mm/h)
I V grass: 24. # Infiltration capacity of vegetated soil (mm/h)
pc grass: 0.43 # Soil porosity (None)
fc grass: 0.56 # Soil saturation degree at field capacity (None)
sc_grass: 0.33 # Soil saturation degree at stomatal closure (None)
wp grass: 0.13 # Soil saturation degree at wilting point (None)
hgw grass: 0.1 # Soil saturation degree at hygroscopic point (None)
beta_grass: 13.8 # Deep percolation constant = 2*b+4 where b is water retention
parameter
# Shrub
VEGTYPE shrub: 1 # Integer value to infer Vegetation Type
intercept cap shrub: 1.5 # Full canopy interception capacity (mm)
zr_shrub: 0.5 # Root depth (m)
I B shrub: 20. # Infiltration capacity of bare soil (mm/h)
I_V_shrub: 40. # Infiltration capacity of vegetated soil (mm/h)
pc_shrub: 0.43 # Soil porosity (None)
fc shrub: 0.56 # Soil saturation degree at field capacity (None)
sc shrub: 0.24 # Soil saturation degree at stomatal closure (None)
wp_shrub: 0.13 # Soil saturation degree at wilting point (None)
hgw shrub: 0.1 # Soil saturation degree at hygroscopic point (None)
beta shrub: 13.8 # Deep percolation constant = 2*b+4 where b is water retention
parameter
# Tree
VEGTYPE tree: 2 # Integer value to infer Vegetation Type
intercept_cap_tree: 2. # Full canopy interception capacity (mm)
zr tree: 1.3 # Root depth (m)
I B tree: 20. # Infiltration capacity of bare soil (mm/h)
I V tree: 40. # Infiltration capacity of vegetated soil (mm/h)
pc tree: 0.43 # Soil porosity (None)
fc tree: 0.56 # Soil saturation degree at field capacity (None)
sc_tree: 0.22 # Soil saturation degree at stomatal closure (None)
wp tree: 0.15 # Soil saturation degree at wilting point (None)
hgw tree: 0.1 # Soil saturation degree at hygroscopic point (None)
beta tree: 13.8 # Deep percolation constant = 2*b+4 where b is water retention
parameter
# Bare Soil
VEGTYPE bare: 3 # Integer value to infer Vegetation Type
intercept_cap_bare: 1. # Full canopy interception capacity (mm)
zr bare: 0.15 # Root depth (m)
I B bare: 20. # Infiltration capacity of bare soil (mm/h)
I_V_bare: 20. # Infiltration capacity of vegetated soil (mm/h)
pc_bare: 0.43 # Soil porosity (None)
fc bare: 0.56 # Soil saturation degree at field capacity (None)
sc bare: 0.33 # Soil saturation degree at stomatal closure (None)
```

# Grass

```
wp bare: 0.13 # Soil saturation degree at wilting point (None)
hgw_bare: 0.1 # Soil saturation degree at hygroscopic point (None)
beta_bare: 13.8 # Deep percolation constant
## Vegetation Dynamics:
Blive init: 102.
Bdead init: 450.
PET growth threshold: 3.8 # PET threshold for growing season (mm/d)
PET_dormancy_threshold: 6.8 # PET threshold for dormant season (mm/d)
Tdmax: 10. # Constant for dead biomass loss adjustment (mm/d)
w: 0.55 # Conversion factor of CO2 to dry biomass (Kg DM/Kg CO2)
# Grass
WUE grass: 0.01 # Water use efficiency KgCO2Kg-1H2O
cb grass: 0.0047 # Specific leaf area for green/live biomass (m2 leaf q-1 DM)
cd grass: 0.009 # Specific leaf area for dead biomass (m2 leaf g-1 DM)
ksg grass: 0.012 # Senescence coefficient of green/live biomass (d-1)
kdd grass: 0.013 # Decay coefficient of aboveground dead biomass (d-1)
kws grass: 0.02 # Maximum drought induced foliage loss rate (d-1)
LAI_max_grass: 2. # Maximum leaf area index (m2/m2)
LAIR_max_grass: 2.88 # Reference leaf area index (m2/m2)
# Shrub
WUE shrub: 0.0025 # Water use efficiency KgCO2Kg-1H2O
cb shrub: 0.004 # Specific leaf area for green/live biomass (m2 leaf g-1 DM)
cd shrub: 0.01 # Specific leaf area for dead biomass (m2 leaf q-1 DM)
ksg shrub: 0.002 # Senescence coefficient of green/live biomass (d-1)
kdd_shrub: 0.013 # Decay coefficient of aboveground dead biomass (d-1)
kws shrub: 0.02 # Maximum drought induced foliage loss rate (d-1)
LAI max shrub: 2. # Maximum leaf area index (m2/m2)
LAIR_max_shrub: 2. # Reference leaf area index (m2/m2)
# Tree
WUE tree: 0.0045 # Water use efficiency KgCO2Kg-1H2O
cb tree: 0.004 # Specific leaf area for green/live biomass (m2 leaf g-1 DM)
cd_tree: 0.01 # Specific leaf area for dead biomass (m2 leaf g-1 DM)
ksg tree: 0.002 # Senescence coefficient of green/live biomass (d-1)
kdd tree: 0.013 # Decay coefficient of aboveground dead biomass (d-1)
kws tree: 0.01 # Maximum drought induced foliage loss rate (d-1)
LAI max tree: 4. # Maximum leaf area index (m2/m2)
LAIR max tree: 4. # Reference leaf area index (m2/m2)
# Bare
WUE bare: 0.01 # Water use efficiency KgCO2Kg-1H2O
cb bare: 0.0047 # Specific leaf area for green/live biomass (m2 leaf g-1 DM)
cd bare: 0.009 # Specific leaf area for dead biomass (m2 leaf g-1 DM)
ksg_bare: 0.012 # Senescence coefficient of green/live biomass (d-1)
kdd_bare: 0.013 # Decay coefficient of aboveground dead biomass (d-1)
kws bare: 0.02 # Maximum drought induced foliage loss rate (d-1)
LAI max bare: 0.01 # Maximum leaf area index (m2/m2)
```

```
LAIR max bare: 0.01 # Reference leaf area index (m2/m2)
## Cellular Automaton Vegetation:
# Grass
Pemaxg: 0.35 # Maximal establishment probability
ING: 2 # Parameter to define allelopathic effect on grass from cresotebush
ThetaGrass: 0.62 # Drought resistant threshold
PmbGrass: 0.05
                             # Background mortality probability
# Shrub
Pemaxsh: 0.2 # Maximal establishment probability
ThetaShrub: 0.78 # Drought resistant threshold
PmbShrub: 0.03 # Background mortality probability
tpmaxShrub: 600 # Maximum age (yr)
# Tree
Pemaxtr: 0.25 # Maximal establishment probability
ThetaTree: 0.72 # Drought resistant threshold
PmbTree: 0.01 # Background mortality probability
tpmaxTree: 350 # Maximum age (yr)
# ShrubSeedling
ThetaShrubSeedling: 0.64 # Drought resistant threshold
PmbShrubSeedling: 0.03 # Background mortality probability
tpmaxShrubSeedling: 18 # Maximum age (yr)
# TreeSeedling
ThetaTreeSeedling: 0.64 # Drought resistant threshold
PmbTreeSeedling: 0.03 # Background mortality probability
tpmaxTreeSeedling: 18 # Maximum age (yr)
```

#### Script S6. Code to run a surface runoff model in Landlab. The file

'Watershed\_DEM.asc' could be any Esri ASCII-formatted file giving rasterised topographic data on which to run this Landlab model. However, the output seen in Figure 15 of the main manuscript and described in the main text makes use of a DEM of Spring Creek, CO, USA. Availability of this data is described in Section 8 of the main text.

```
1
     from landlab.io import read esri ascii
2
     from landlab.components import SinkFiller, OverlandFlow
3
     (mg, z) = read_esri_ascii('Watershed_DEM.asc',
                               name='topographic elevation')
     mg.set_watershed_boundary_condition(z)
     sf = SinkFiller(mg, routing='D4', apply_slope=True, fill_slope=1.e-5)
     sf.fill pits()
    of = OverlandFlow(mg, steep slopes=True)
    elapsed time = 0.
9
10
    storm duration = 3600. # in seconds
    model_run_time = 84600. # in seconds
11
     starting precip = 25. # in mm/h
12
13
     while elapsed time < model run time:
14
         if elapsed_time < storm_duration:</pre>
             of.rainfall_intensity = starting_precip * 2.7778e-7 # mm/h to m/s
15
16
         else:
17
             of.rainfall intensity = 0.
18
         of.run one step() # this component can select its own timestep
         elapsed time += of.dt
19
```