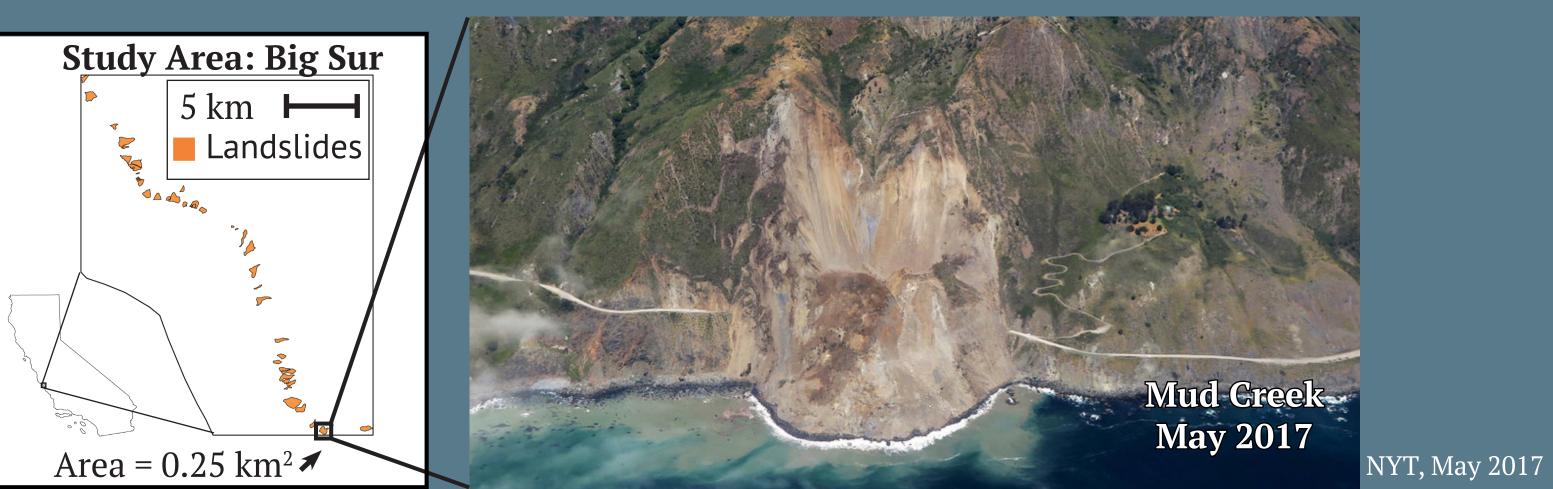
INVESTIGATING THE FATE OF CREEPING LANDSLIDES WITH COMPLEX NETWORK THEORY

Vrinda Desai¹, Farnaz Fazelpour¹, Alexander Handwerger^{2,3}, Karen E. Daniels¹ ¹Department of Physics, North Carolina State University, Raleigh, NC ² Joint Institute for Regional Earth System Science and Engineering, University of California, Los Angeles, CA ³Jet Propulsion Laboratory, California Institute of Technology, Pasadena, CA *Contact: vddesai@ncsu.edu*

Global warming has contributed to an increased risk of California drought due to rising temperatures [1]. Combined with an increased likelihood of wet years in California, slow-moving landslides are more likely to fail due to saturated soils.

Why a Network?

- Is useful for datasets that are connected through space and time
- Can help mine large datasets for consistent patterns using well-developed toolboxes
- A complex network:
- Contains a set of nodes connected by edges that can be weighted and/or directed
- Can represent temporal dynamics via a multilayer network of spatial slices



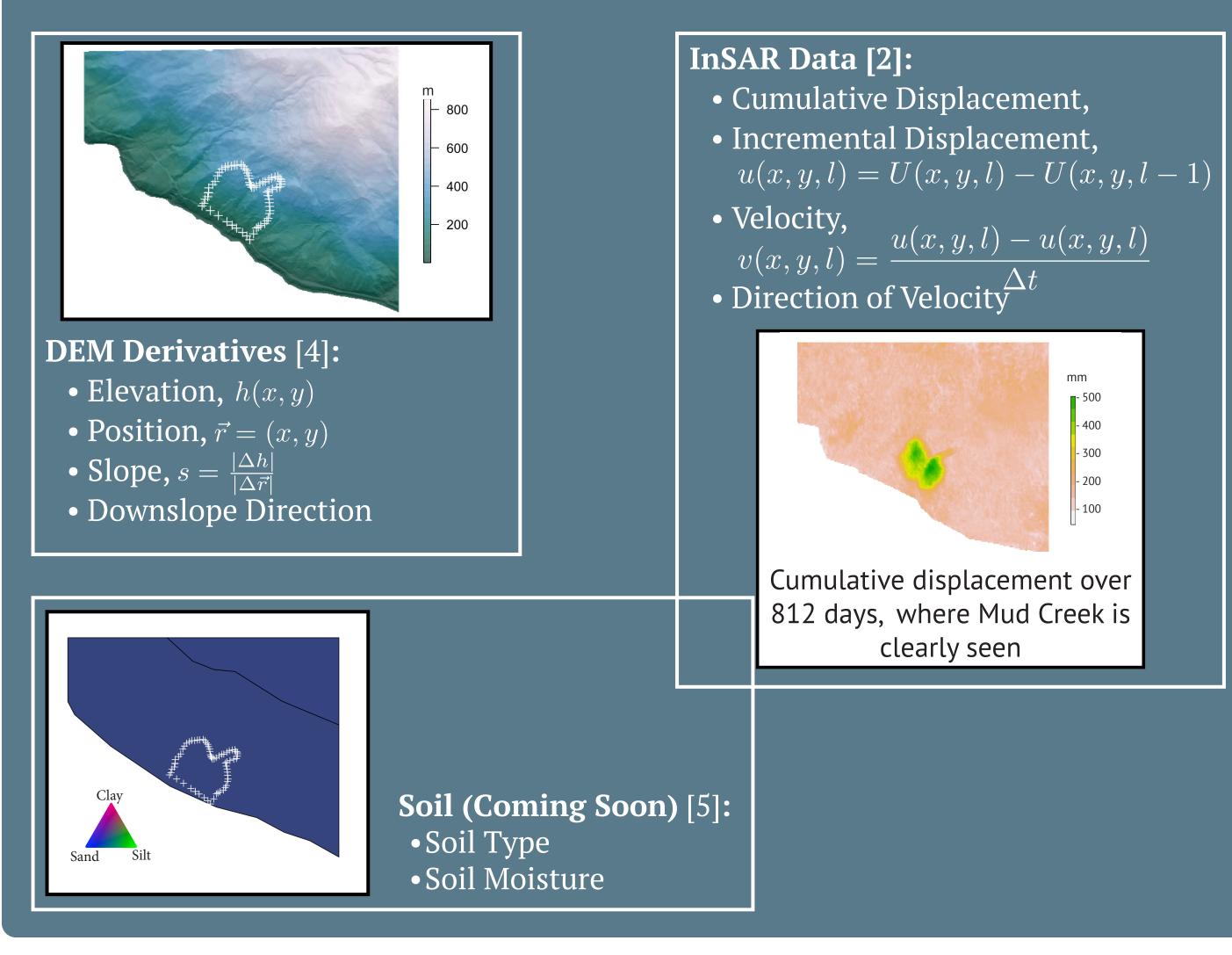


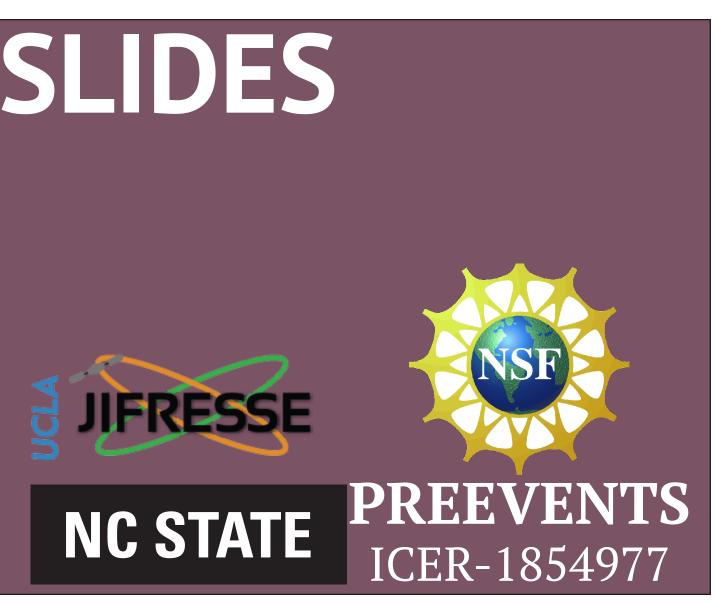


Digital **Elevation** Model

What is Community Detection? • Identifies clusters of nodes that are more strongly connected to each other than they are to the rest of the network

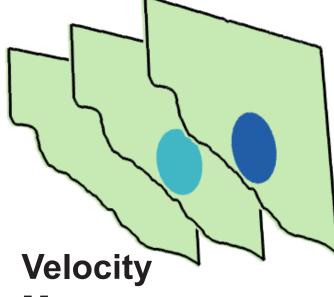
Selecting Variables to Use in Forecasting Failure



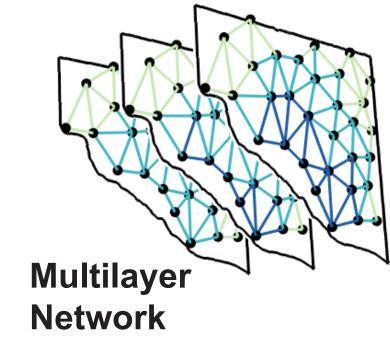


Rainfall-Induced Landslides

Network Science Framework



Maps



Low γ Low ω

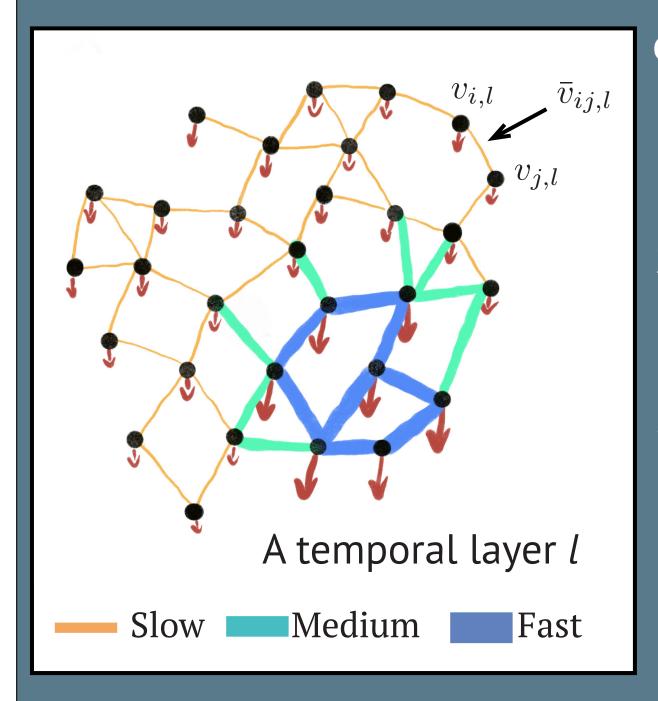
- 400

- 100

Which Technique do we use? GenLouvain - optimizing modularity [3]

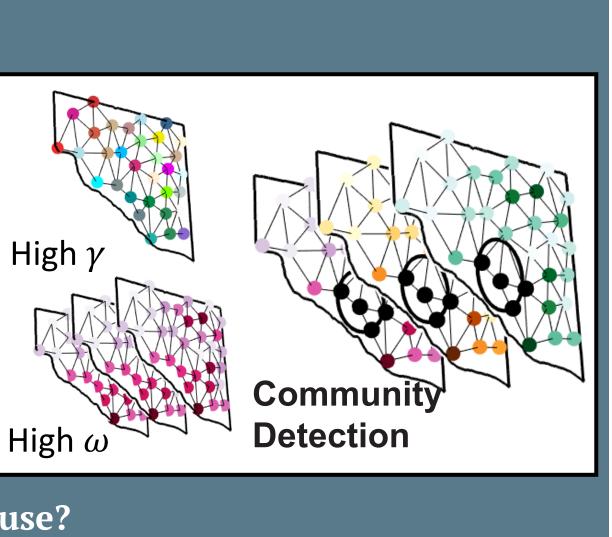
Weights for Multilayer Network

To incorporate the current state of rheology as well as susceptibility of a hillslope, we used velocity, derived from InSAR, and slope from DEM. There are a total of 62 layers from the InSAR data that is taken every 6,12, or 24 days [2].



- where

Using satellite InSAR data, we apply techniques from network science to develop methods to investigate spatio-temporal patterns to predict the sudden transition from gradual deformation to runaway acceleration and catastrophic failure.



Calculating Edge Weights

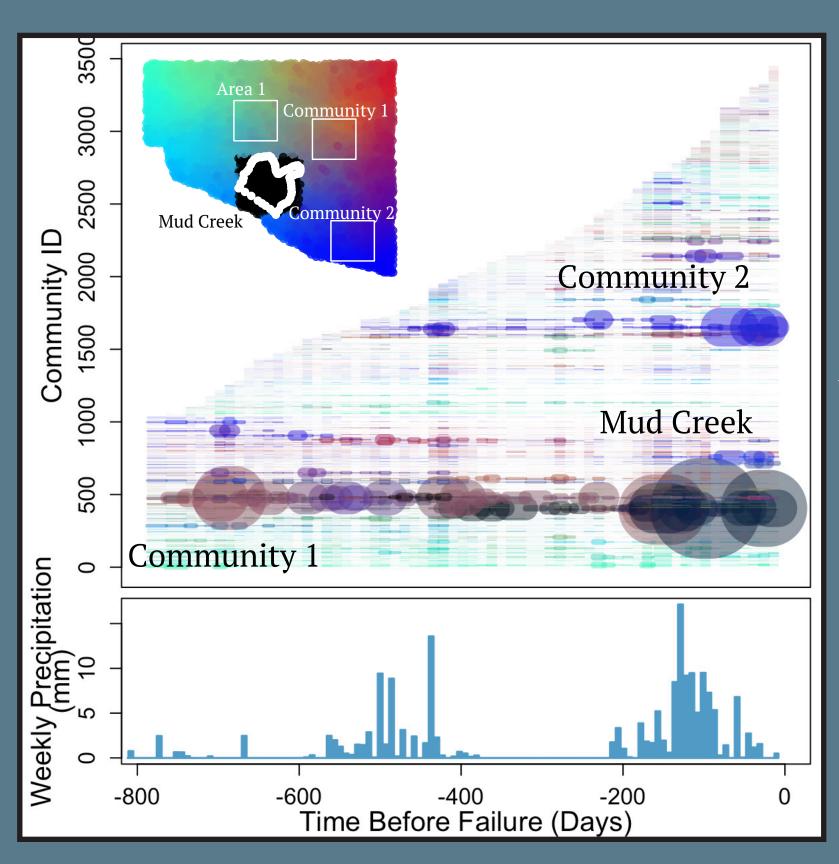
. We account for the uneven time intervals by calculating velocity for any node *i*: $v_{i,l} = \frac{u_{i,l}}{\Delta t}$

2. We assign average velocity, $\bar{v}_{ij,l} = \frac{|v_{i,l} + v_{j,l}|}{2}$, as edge weights for any 2 connected nodes

3. We account for both the local slope and recent velocity by defining the adjacency matrix *M*, as an undirected, weighted multilayer matrix,

> If nodes i and j are connected, otherwise,

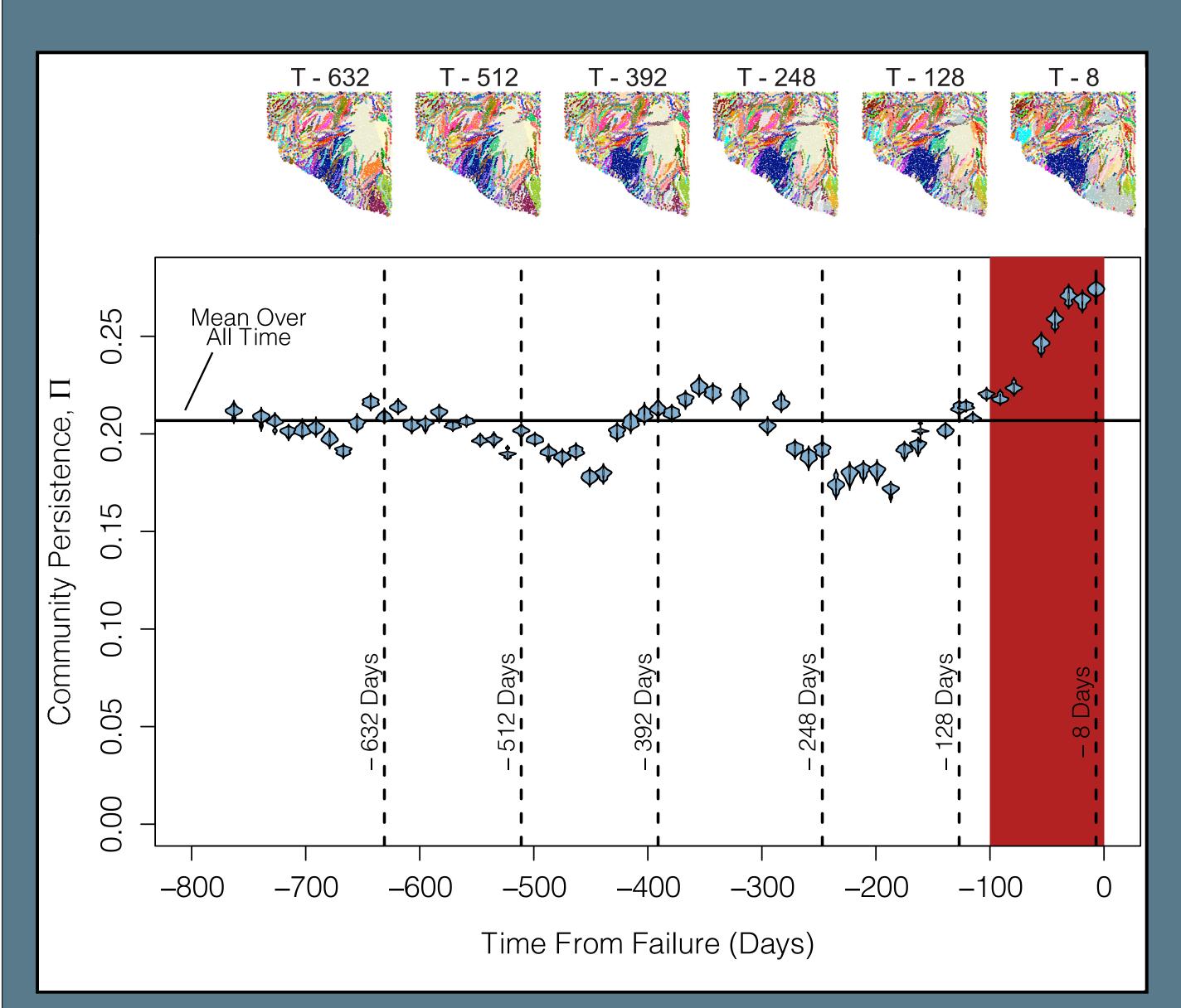
Forecasting Landslides Using Community Detection



Community Persistence: Measures the stability of the nodal composition for each community in relation to the community's size

 $\Pi_l = \frac{1}{N} \sum_c \frac{|c_{l-1} \cap c_l|}{n_{c,l}}$

N: Total number of nodes $n_{c,l}$: Number of nodes in community c at layer l $|c_{l-1} \cap c_l|$: Number of nodes present in community *c* in both layers *l* and *l-1*



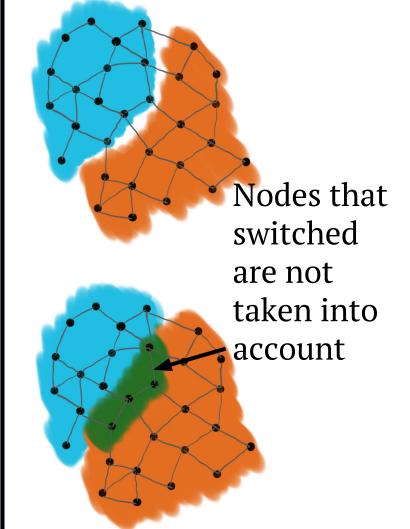
- failure in the weeks leading up. as stable vs unstable.
- [2] Handwerger et al., Scientific Reports, 9:1569 (2019) [3] <u>http://netwiki.amath.unc.edu (open-source)</u>

Key Observations:

- GenLouvain consistently identifies Mud Creek as a community
- Several other communities are also identified intermittently, but appear less consistently.

We selected 4 areas for comparison: • 'Area 1' has similar topography to Mud Creek (at higher elevation), but was never identified as a community • 'Community 1' & 'Community

2' were repeatedly identified as communities, but did not fail



• Using community detection, we detect patterns that allows us to forecast Mud Creek's • We are able to distinguish the three creeping landslides found within this study area