Sebastian Busby GEOG 593 - FALL 2018

Project Abstract

Topography plays an important role in influencing the behavior of wildfires in forested landscapes. Unlike other factors that change over time (e.g. climate, weather, fuel aridity), topography is relatively static and therefore easier to quantify as an influential factor. Within recently field sampled wildfire perimeters on Mt. Adams, WA and Mt. Jefferson, OR, spatially derived burn severity (RdNBR) was compared to five measures of topography derived from a Digital Elevation Model (DEM) at a spatial scale of 30m. These topographic variables included elevation, slope, Heat Load Index (HLI), Topographic Wetness Index (TWI), and Topographic Roughness Index (TRI). Cell by cell correlation between variables was explored using a correlation matrix in R. Topographic variables were unsurprisingly at least moderately collinear. To account for geographic relationships, a limited number of Geographically Weighted Regression (GWR) models were developed using ArcGIS. Among the two field sites, the TWI GWR model best fit the data and had the highest R2 values (0.92, 0.83). TWI is a steady-state wetness index that indicates where water is most likely to be funneled to and pool in. Intuitively, forest stands with a high moisture content during a fire are less likely to catch fire or burn at high severities. Within these field sites, the TWI appears to be the best topographic predictor of wildfire burn severity.

Keywords: Wildfire, Topography, Burn Severity, Heat Load Index, Topographic Roughness Index, Topographic Wetness Index, Central Cascades, Vegetation Mortality, Forest Regeneration, RdNBR.

Can Wildfire Burn Severity be Explained by Measures of Topography?



Sebastian Busby - DTA Fall 2018

Why do some burned area have nearby live trees, and others don't?



→ No nearby live trees (seed source) = poor post-fire tree regeneration

Drivers of Wildfire Burn Severity (Conceptual Understanding)

1. Pre-fire Climate

a. Winter snowpack, Spring/Summer temperature and precipitation

2. Weather Conditions During Fire

a. Temperature, precipitation, and active winds

3. Forest Composition

a. Certain vegetation is more or less flammable - inhibit or spread fire

4. Forest Fuel Density

a. How dense is vegetation pre-fire; generally more fuel = more fire

5. Topographic Factors

a. Static landscape properties that influence all of the above factors



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Flat topography: no wind

Flat topography; wind

Hillslope and wind

Topography influences wildfire behavior through:

Influence of Topography

- 1. Convective funneling of hot air masses
- 2. Buffer or funnel for wind
- 3. Alters the flow and accumulation of precipitation runoff
- 4. Alters soil and vegetation moisture (slope, runoff)
- 5. Aspect & Slope alter solar radiation potential (site wetness)



Research Methods

For my field research areas on Mt. Adams and Mt. Jefferson:

- 1. Derive wildfire burn severity (RDNBR) as my dependant variable from MTBS data (30m resolution)
- 2. Derive topographic explanatory variables from a DEM that may conceptually explain burn severity
 - a. Elevation
 - b. Slope
 - c. Heat Load Index (HLI)
 - d. Topographic Wetness Index (TWI)
 - e. Topographic Ruggedness Index (TRI)
- 3. Mask all layers to burn perimeters
- 4. Explore variable correlation using a correlation matrix in R
- 5. Develop Geographically Weighted Regression (GWR) models in ArcGIS

Relativized differenced Normalized Burn Ratio (RDNBR) Wildfire Burn Severity metric



Elevation (m)



Slope (°)



Heat Load Index (HLI)



(McCune and Keon 2002)

Topographic Wetness Index (TWI)



(Beven and Kirkby 1979)

Topographic Roughness Index (TRI) 10x10 window (9ha)





Correlation Matrix

- Elevation has a weak correlation with RDNBR on Mt. Adams, not Mt. Jefferson
- No other variables seem to have a correlation with RDNBR
- Slope and TRI are highly collinear
- Other topographic variables have moderate collinearity
- → Collinearity not surprising, all measures of topography



Geographically Weighted Regression

Issues:

- GWR would not run with some of the predictor variables; Elevation, HLI, Slope
- Poor performance of the multiple regression models is likely due to collinearity of variables

| Mt. Adams | Performance | Response | Predictor(s) | Sigma | AICc | R2 | R2 Adj | Residual Squares |
|---------------|-------------|----------|--------------|-------|---------|------|--------|-------------------------|
| Model 1 | #2 | RDNBR | TRI | 150 | 741,677 | 0.84 | 0.83 | 1,253,251,222 |
| Model 2 | #1 | RDNBR | TWI | 115 | 715,833 | 0.92 | 0.9 | 624,967,802 |
| Model 3 | #3 | RDNBR | Slope | 163 | 750,778 | 0.81 | 0.8 | 1,487,800,704 |
| Model 4 | #4 | RDNBR | TWI + TRI | 167 | 754,106 | 0.8 | 0.79 | 1,562,712,827 |
| | | | | | | | | |
| Mt. Jefferson | Performance | Response | Predictor(s) | Sigma | AICc | R2 | R2 Adj | Residual Squares |
| Model 1 | #2 | RDNBR | TRI | 207 | 353,019 | 0.58 | 0.58 | 1,115,989,903 |
| Model 2 | #1 | RDNBR | TWI | 139 | 334,100 | 0.83 | 0.81 | 439,417,900 |
| Model 3 | #3 | RDNBR | TWI + TRI | 213 | 354,415 | 0.56 | 0.55 | 1,170,892,586 |

Results:

• Across both field sites, the Topographic Wetness Index (TWI) produced the best GWR model and explained the most variance (R2) in wildfire burn severity.

Do These Results Make Sense? Yes!

- TWI is a steady-state wetness index.
- Shows where water topographically is most likely to be funneled to and pool in.
- Forests stands with a high moisture content during a fire are much less likely to burn!
- Therefor high TWI values should relate to a lower RDNBR values.
- Take results with a grain of salt, non-topographic factors are important and can overpower influence of TWI.



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Datasets

- 1. Burn Severity (RdNBR): Monitoring Trends in Burn Severity (MTBS)
- 2. 10m DEM: University of Washington (WA); Oregon Spatial Data Library (OR)