

Predicting Post-wildfire Debris Flow Occurrence in Western USA

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Data

❖ United States Geological Survey (USGS)

- Open-File Report 2005-1218

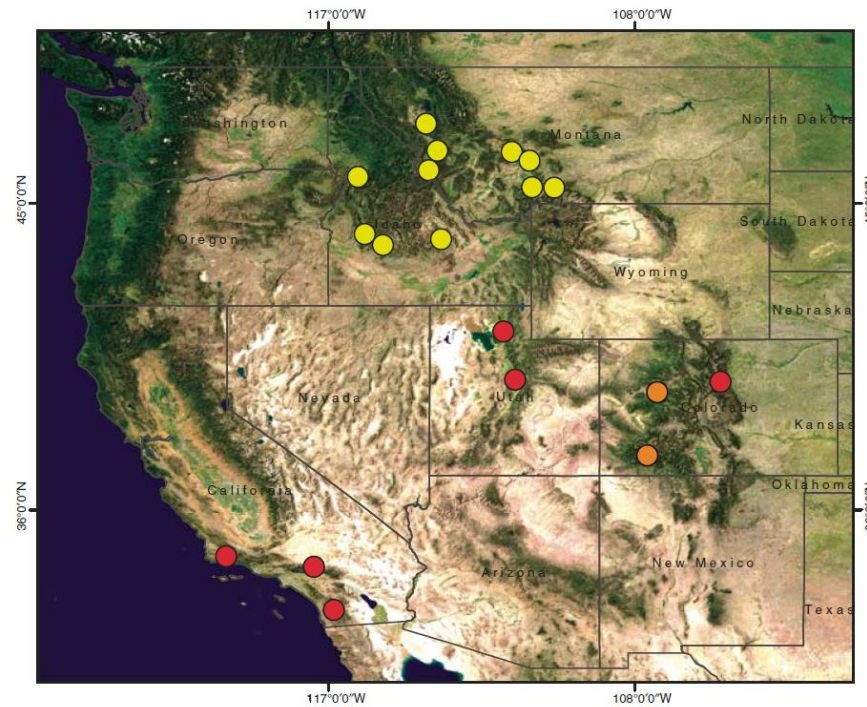
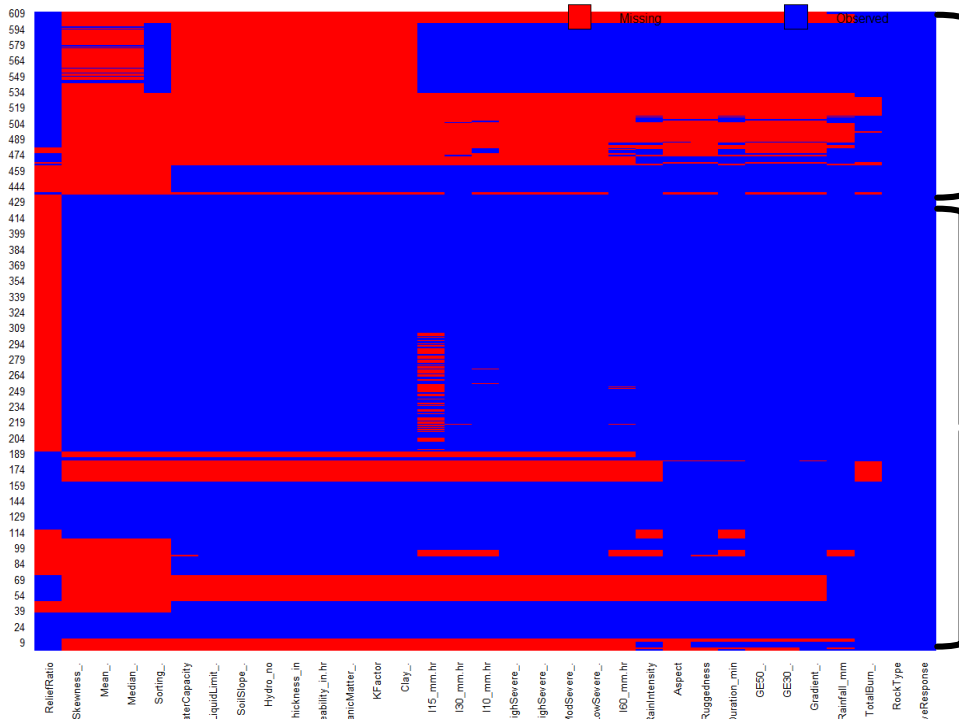


Figure 1. Map showing locations of basins used to develop models for the probability of debris-flow generation (yellow dots), for estimates of debris-flow volume (red dots), or both models (orange dots).

Data



California (144 samples)

Intermountain West (465 samples)

❖ 33 predictors

❖ 609 sample points

Pre-processing Steps

Missing
data

Predictor
degeneracy

Skewness and
scaling

Dummy
variables

Pairwise
correlation

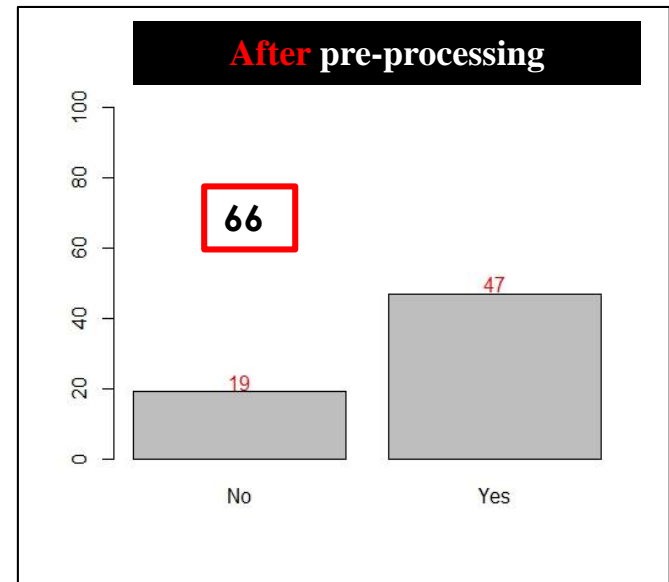
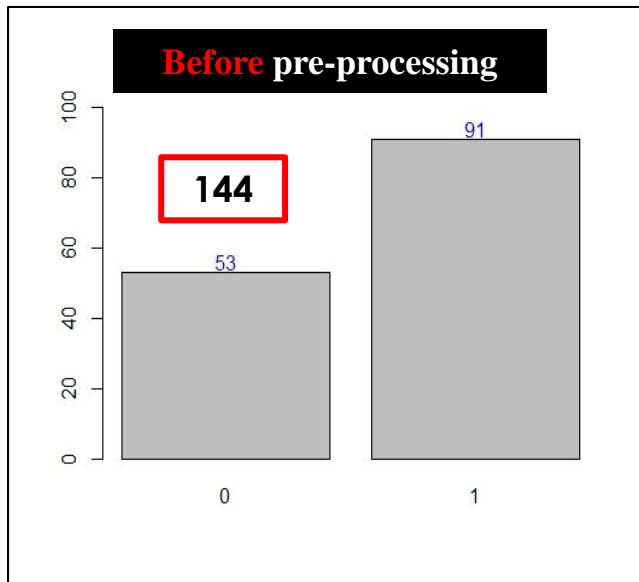
Different data explorations

- Southern California (SC)
- Intermountain (Im)

Southern California

Data

Southern California



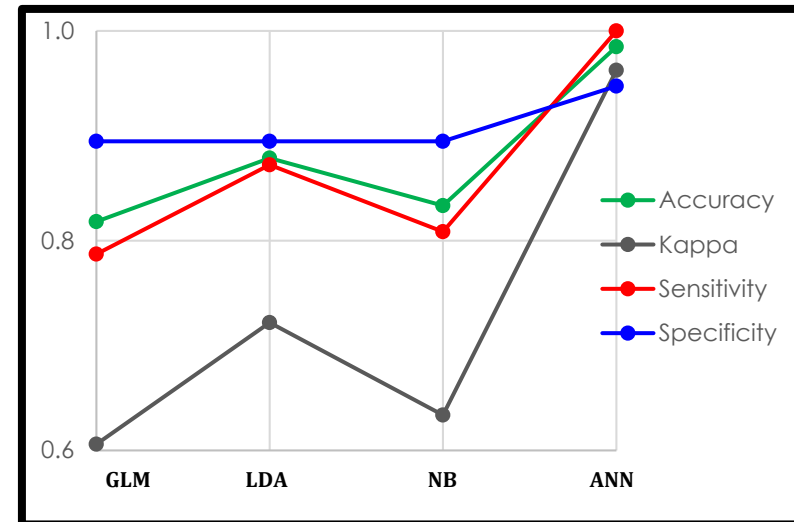
10 fold cross validation repeated 10 times

Results

Southern California

Training Metrics

| MODEL | ACC. | KAPPA | SENS. | SPEC. |
|------------------------------------|-------|-------|-------|-------|
| Logistic Regression (GLM) | 0.818 | 0.606 | 0.787 | 0.895 |
| Linear Discriminant Analysis (LDA) | 0.879 | 0.722 | 0.872 | 0.895 |
| Naïve Bayes (NB) | 0.833 | 0.634 | 0.809 | 0.895 |
| Averaged Neural Network (ANN) | 0.985 | 0.962 | 1.000 | 0.947 |

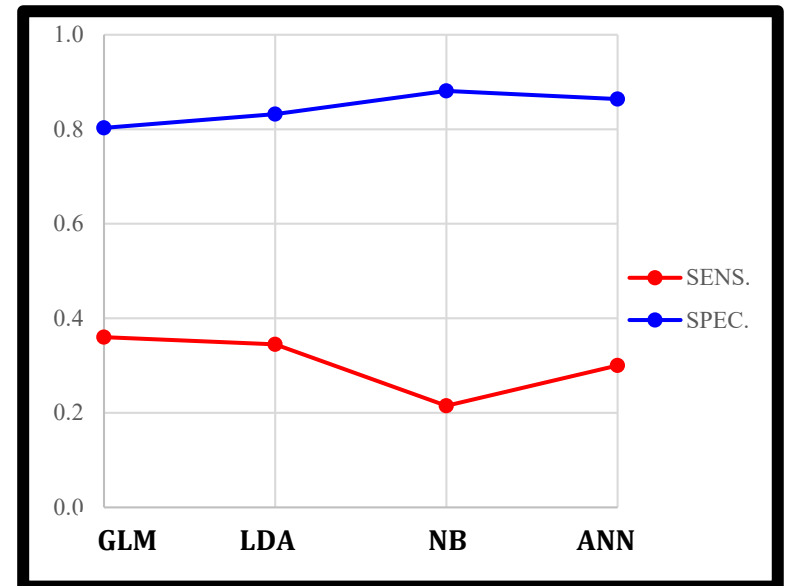


Results

Southern California

Resampled Metrics - 10 fold CV

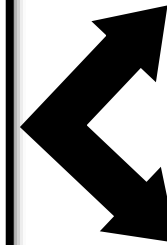
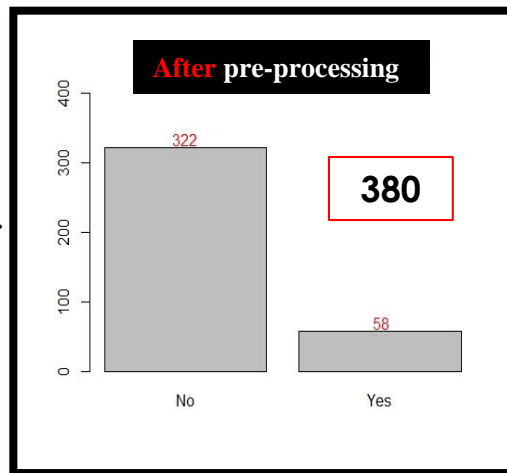
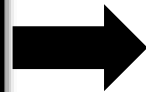
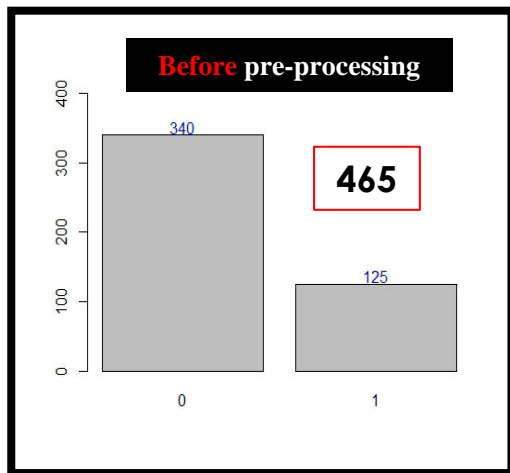
| MODEL | SENS. | SPEC. |
|------------------------------------|-------|-------|
| Logistic Regression (GLM) | 0.360 | 0.803 |
| Linear Discriminant Analysis (LDA) | 0.345 | 0.832 |
| Naïve Bayes (NB) | 0.215 | 0.881 |
| Averaged Neural Network (ANN) | 0.300 | 0.864 |



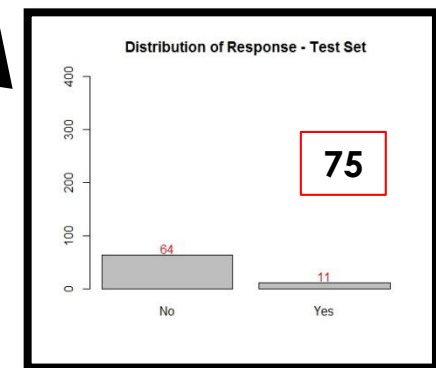
Intermountain

Data

Intermountain



80%



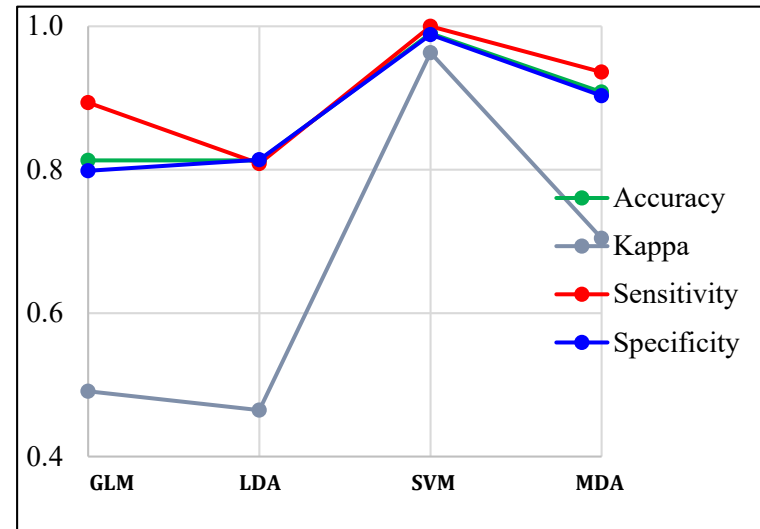
20%

Results

Intermountain

Training Metrics

| MODEL | ACC. | KAPPA | SENS. | SPEC. |
|-------------------------------------|-------|-------|-------|-------|
| Logistic Regression (GLM) | 0.813 | 0.491 | 0.894 | 0.798 |
| Linear Discriminant Analysis (LDA) | 0.813 | 0.465 | 0.809 | 0.814 |
| Support Vector Machine (SVM) | 0.990 | 0.963 | 1.000 | 0.988 |
| Mixture Discriminant Analysis (MDA) | 0.908 | 0.704 | 0.936 | 0.903 |

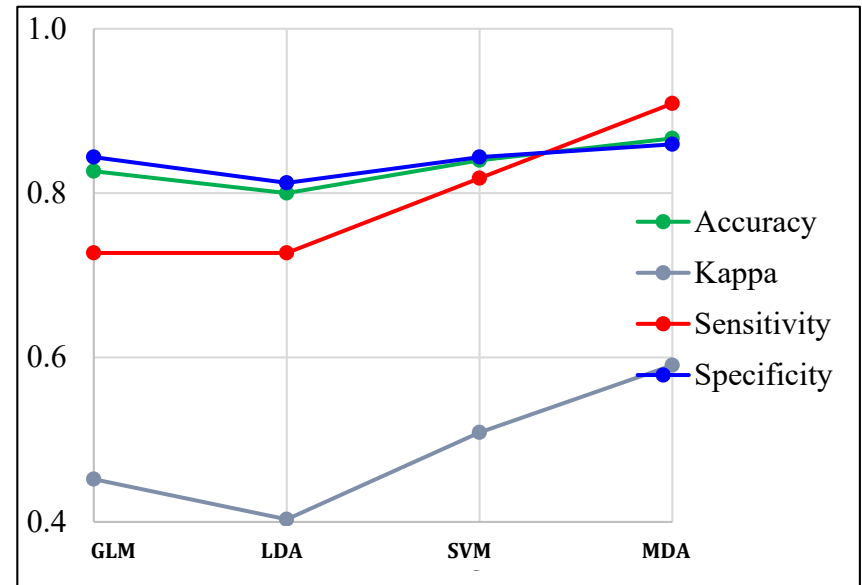


Results

Intermountain

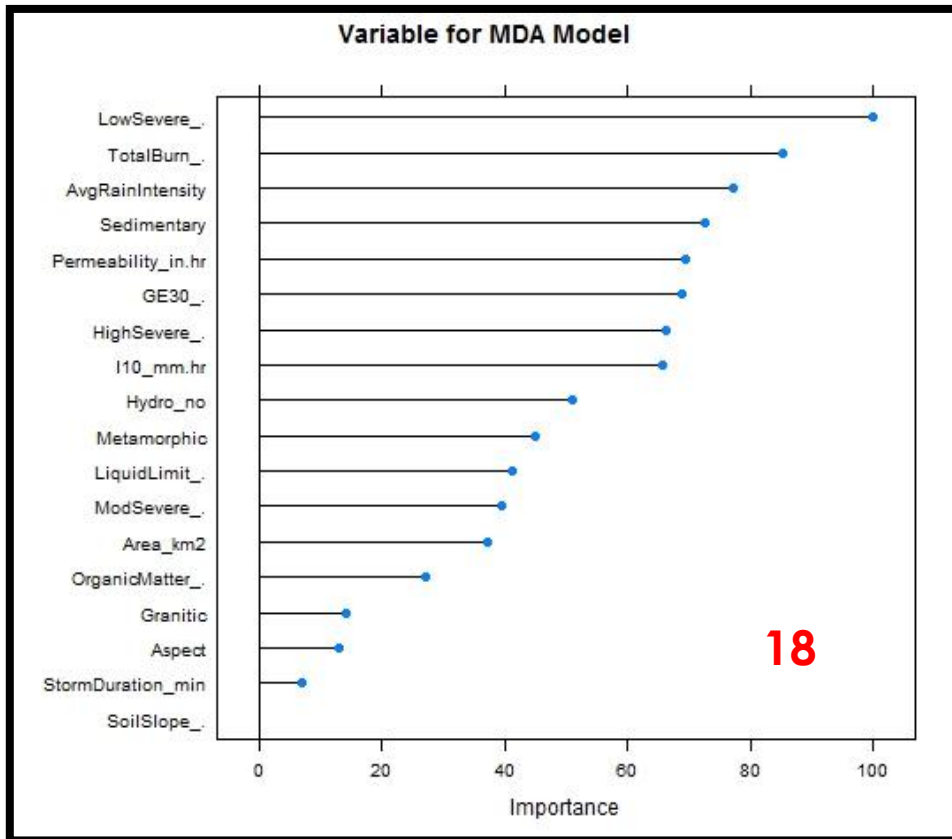
Validation Metrics

| MODEL | ACC. | KAPPA | SENS. | SPEC. |
|-------------------------------------|-------|-------|-------|-------|
| Logistic Regression (GLM) | 0.827 | 0.452 | 0.727 | 0.844 |
| Linear Discriminant Analysis (LDA) | 0.800 | 0.403 | 0.727 | 0.813 |
| Support Vector Machine (SVM) | 0.840 | 0.509 | 0.818 | 0.844 |
| Mixture Discriminant Analysis (MDA) | 0.867 | 0.591 | 0.909 | 0.859 |



Results

Intermountain



Conclusions and Future Work

- ❑ Nonlinear models performing better than linear models suggest an underlying nonlinear relationship between predictors and response variable.
- ❑ California data too scanty to glean a trend from it. Further data collection and model development required to improve predictions.
- ❑ Intermountain data performs better with a sensitivity of 91% for nonlinear MDA model in comparison with 44% of existing linear model (Cannon et al, 2010).
- ❑ Future work aims at:
 - ✓ Testing model validity with independent data obtained from Nevada BLM.
 - ✓ Developing linear and nonlinear volumetric models.