# Predictive Modeling for Flood Resiliency

Figure 1

Figure 2

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> The purpose of this project is to develop a predictive model that predicts the probability that an area will be inundated with flood water. The model has been trained and validated on Calgary, a Canadian City in Alberta, and then used to predict for a comparable city, Denver Colorado. This report will explain the planning motivation as well as illustrate the environment features, model performance, and prediction results.

### **Motivation**

The motivation is to strengthen flood resiliency. Climate change is expected to affect water levels of not only coastal but also inland waters, as changing precipitation and outdated infrastructure have already led to alarming cases in places such as Calgary. While research and planning efforts have not been prevalent for riverine flooding, local governments suffer greatly from unexpected floods. Establishing predictive а model will be necessary and helpful to for governments to designate flood zones to allocate prevention resources.

Through exploratory analysis (Figure 1), we recognize that occurrences of inundation are related to environmental features. The distributions for four of them are mapped per 500 x 500 meters grid areas (Figure 2).

# **Environment Features**

#### **Exploratory Analysis of Features**



# Map of Four Important Features



Intensity of Human Modification to Land





NDVI (Normalized Difference Vegetation Index)



# **Logistic Regression Model**

#### Model Performance

Receiver Operating Characteristic (ROC) Curve



#### **Confusion Matrix**

	Reference			Accuracy:
Prediction		No Flood	Flood	95% CI: (0.86 No information rate: P-value [ACC>NIR]: Sensitivity: Specificity: Precision:
	No Flood	850	77	
	Flood	41	98	

# Model Validation

Goodness of Fit Metrics from Cross-validation



With the environment features, we predicted them on logit regression. We used stepwise selection to choose features that are most significant to the flood probabilities. The features mentioned above along two slope characters are significantly related to flooding probabilities.

Figure 3

This ROC Curve indicates the goodness of fit of the model. With the area under the curve at 0.8724, our model could distinguish inundation and none inundation area quite well.

#### Table 1

0.8893

0.8358

0.56

0.95

0.70

4.792e-07

(0.8689), (0.9075)

Confusion matrix describes prediction performances on flood and no flood. We hope to not miss any true flood occurrences but also make best use of prevention resources. Our model is able to predict 56% true flood occurrences correctly. Meanwhile, 70% of the predicted floods are correct.

Figure 4

We cross validated it by iteratively testing on 100 random samples. The model predicts relatively well across samples.

#### Sensitivity Rate by Fishnet Grid Area from Spatial Cross-validation



### Figure 5

Our spatial validation method assumes the local spatial process from all other grid areas generalizes to the hold-out.

#### Sensitivity (True Positive rate)

measures the proportion of positives that are correctly identified.

#### Specificity (True Negative rate)

measures the proportion of negatives that are correctly identified.

# **Prediction Result**

