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This research is supporting a CIRNAC lead Climate Risk Index tool. (providing a data layer on flood susceptibility/flood prone areas)

# Prediction and classification of flood susceptibility based on historic record in a large, diverse, and data sparse country

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## Objective of Research Project

- Objective:
  - Create a national dataset that maps flood susceptibility or potentially flood-prone areas based on hydro-geomorphic and meteorological data and flood history
- Why
  - To help understand areas potentially at risk of flooding
  - To identify areas that could be prioritized for high-resolution data collection and flood hazard mapping, support FHIMP
  - This research supports a CIRNAC-led climate risk index tool
- How
  - Using machine learning models of classification and regression algorithms to identify patterns in areas where flooding has historically occurred

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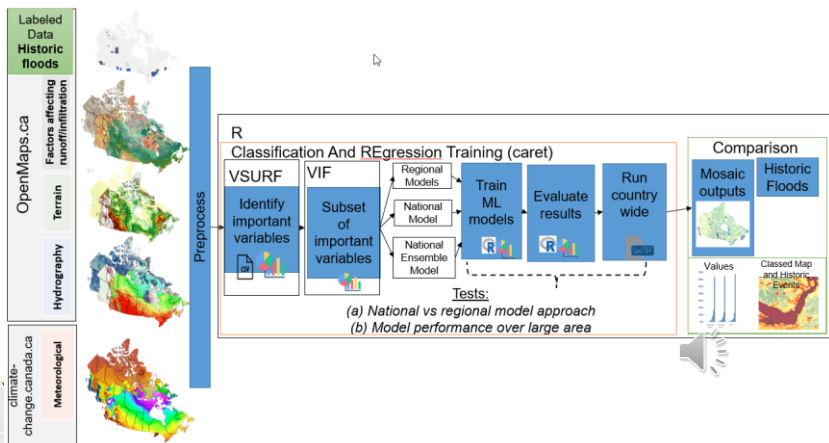
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# Workflow

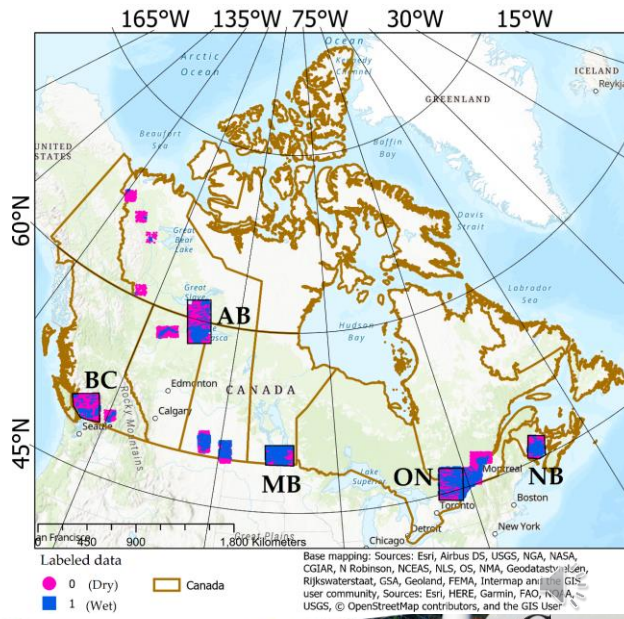
- Develop a method to create a national dataset that maps susceptibility to flooding (or potentially flood prone areas), based on hydro-geomorphology and meteorological data and historic record of flood events

1. Assemble Data
2. Determine 'important' datasets
3. Run several models
4. Evaluate performance
5. Select 'best' performing model
6. Apply model to the rest of the country



# Training data

- Several sites across Canada which have historic record of floods

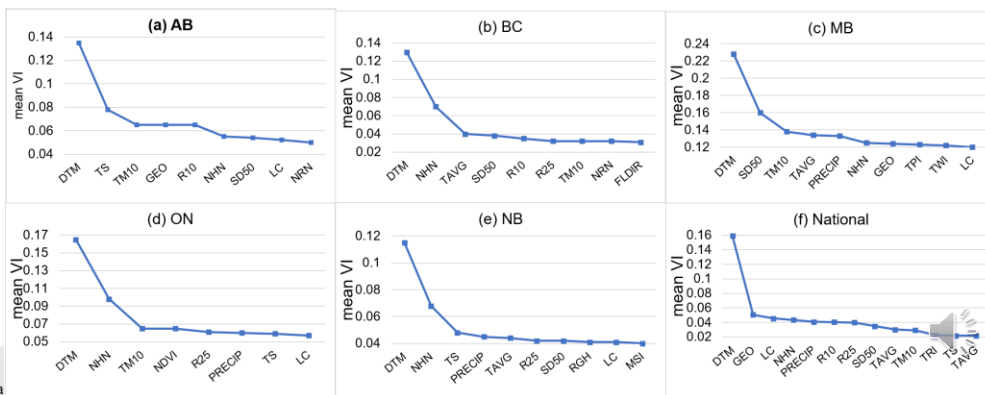


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## 2. Determine the Important datasets

- Through recursive feature elimination (rfe –caret) and Variable Selection using Random forest



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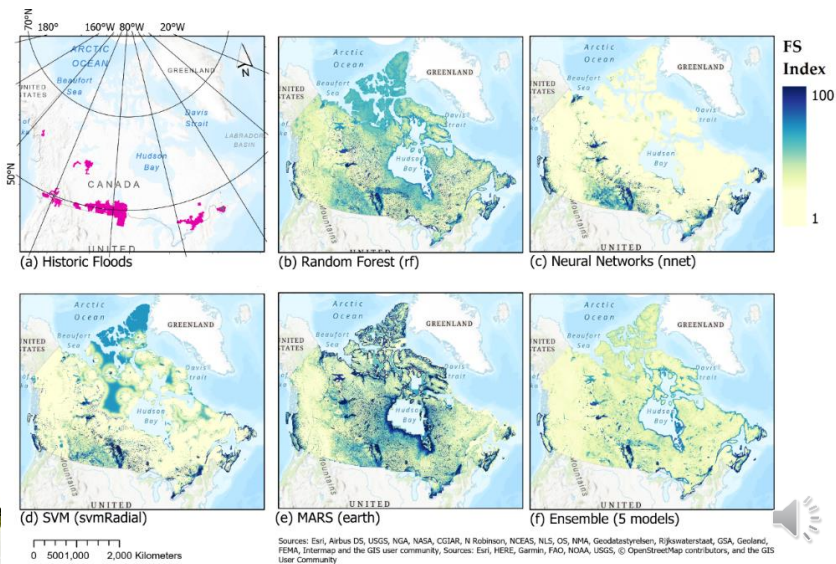
## Single vs multi-regional models (Random Forest)

	Regional Models (parRF) local-variables/national factors						Single Model
	BC	AB	MB	ON	NB	Average	National Model
<b>Accuracy</b>	0.96/0.96	0.94/0.94	0.82/0.82	0.89/0.92	0.99/0.99	0.91/0.93	0.92
<b>Kappa</b>	0.93/0.93	0.88/0.88	0.64/0.64	0.77/0.84	0.97/0.98	0.84/0.85	0.83
<b>Sensitivity</b>	0.95/0.93	0.91/0.91	0.79/0.79	0.91/0.92	0.98/0.98	0.91/0.91	0.91
<b>Specificity</b>	0.98/0.99	0.97/0.97	0.85/0.85	0.85/0.92	0.99/1.0	0.93/0.95	0.9
<b>Precision</b>	0.98/0.98	0.98/0.97	0.81/0.81	0.89/0.92	0.99/1.0	0.91/0.90	0.9
<b>F1</b>	0.95/0.93	0.91/0.91	0.79/0.79	0.91/0.92	0.98/0.98	0.91/0.91	0.91
<b>AUC-ROC</b>	0.97/0.97	0.96/0.96	0.86/0.86	0.92/0.93	0.99/0.99	0.94/0.94	0.97

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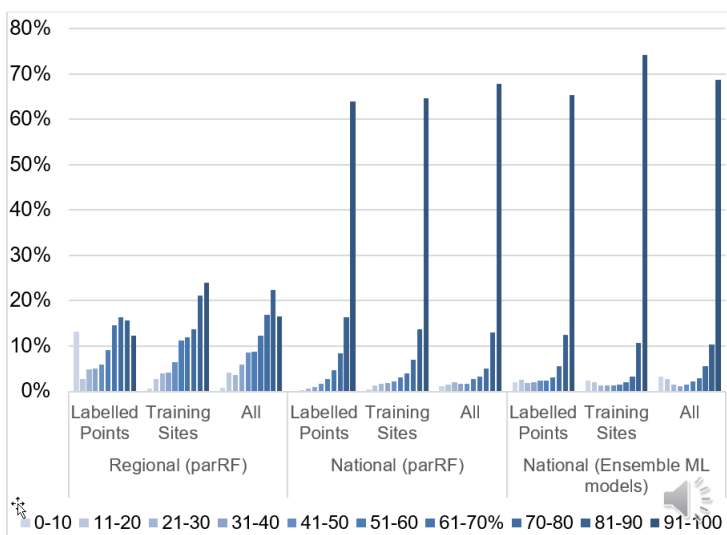
# Single National Model

- RF & MARS somewhat similar
  - Though MARS higher values
- NNET model:
  - good where the training data exists
  - relatively low elsewhere
- SVM:
  - Rings north of training/test sites



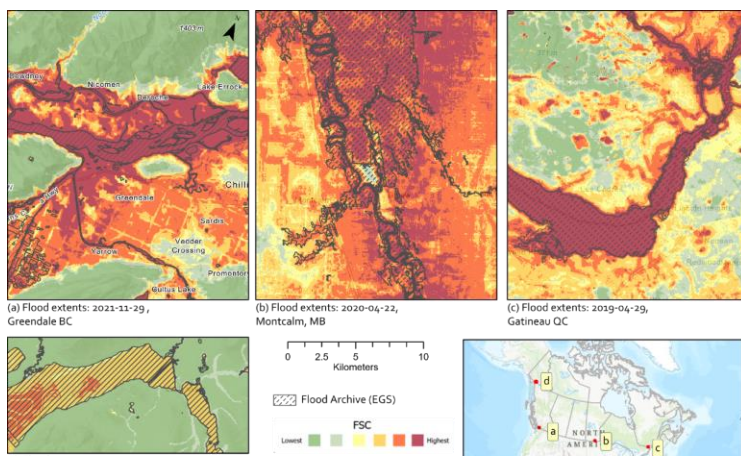
# Comparison of prediction to historic events

- Using multi-regional, single national model and ensemble model approach
  - (i) Labelled points (used in training/test sets), (ii) data in the near-by surrounding sites, and (iii) All data across the country



## Results : Compared to historic events

- The resultant dataset provides the first continuous, national picture of flood susceptibility in Canada, with intended uses of:
  - Supporting identification and priority setting of flood hazard mapping project
  - For flood awareness communication
  - Auxiliary dataset for various climate risk indexes



Sources: Esri, Airbus DS, USGS, NGA, NASA, CGIAR, Robinson, NCEAS, JPL, S. OS, etc. (a) Sources: Esri, HERE, Garmin, FAO, NOAA, etc. (b) Sources: Esri, HERE, Garmin, FAO, NOAA, etc. (c) Sources: Esri, HERE, Garmin, FAO, NOAA, etc. (d) Sources: Esri, HERE, Garmin, FAO, NOAA, etc.

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## Conclusions

- Several challenges exist when trying to develop a flood susceptibility model for a country as large and diverse as Canada:
  - sparse training data
- Findings:
  - single national model produced better predictions than the regional, multi-region model
    - However, when a single ML model was extrapolated across the whole of Canada, there were limitations found in several models, including SVM, NN, MLP and RF.
  - An ensemble approach, which combined 5 models of different categories and combined them using a generalized linear model (GLM), ultimately produced the best flood susceptibility map, in comparison to historic flood maps

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