

# CLARUS Road-Weather Routing for Crash Risk Aversion

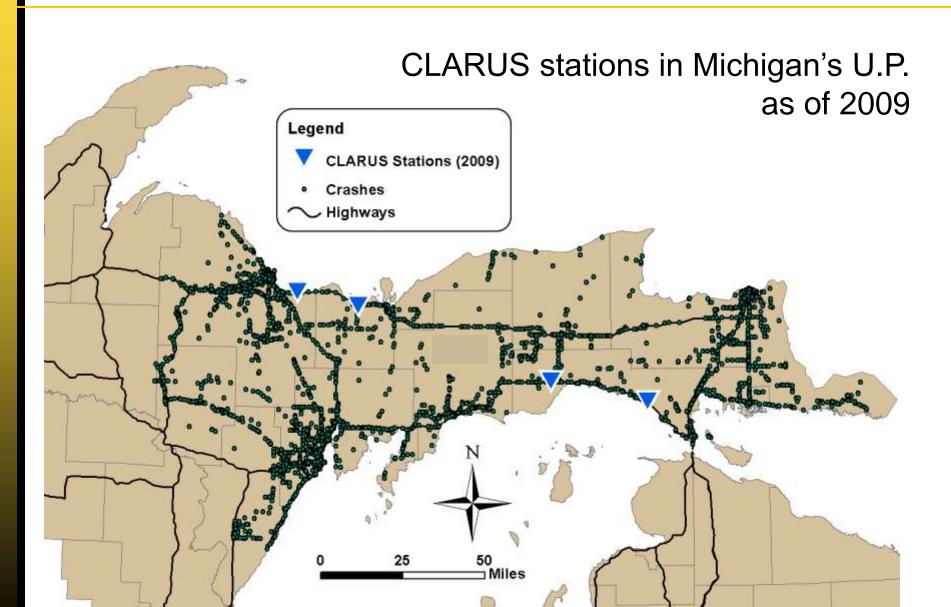
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## **CLARUS Monitoring Stations**





#### **Linking Crashes and Weather**

- A regression model was created
  - Dependent Variable A crash occurring within 50 miles of a weather station during a particular hour.
  - Independent Variables
    - Temperature (Air, Road and Dew Point)
    - Precipitation Types
    - Precipitation Intensities
    - Visibility
    - Wind Speed (Average and Gust)
    - Atmospheric Pressure



#### **Linking Crashes and Weather**

#### First cut: What variables are significant?

Variable	Units	Directionality Tested	Odds of a Crash	p-value
Wind Speed	m/s	Higher Speed	1.023	0.0140
Ice Percent	%	Greater Percent	1.003	0.0288
Heavy Preciptiation Intensity	present	1	1.753	< 0.0001
No Precipitation Present	present	1	0.808	0.0052
Preciptiation as Snow	present	1	2.174	< 0.0001
Atmospheric Pressure	mbar	Higher Pressure	0.993	< 0.0001
Dew Pt Temp	deg C	Higher Temperature	0.984	< 0.0001
Relative Humidity	%	Greater Percent	0.996	< 0.0001
Precipitation Rate	cm/hr	Greater Rate	1.118	0.0478
Visibility	1000m	<b>Greater Visability</b>	0.961	< 0.0001
Air Temperature	deg C	Higher Temperature	0.987	< 0.0001
Dry Road Surface	present	1	1.19	0.0004
Surface Ice Warning	present	1	1.747	0.0099
Surface Ice Watch	present	1	1.321	0.0010
Road Surface Temperature	deg C	Higher Temperature	0.995	< 0.0001
Wind Gust Speed	m/s	Higher Speed	1.042	< 0.0001



#### **Linking Crashes and Weather**

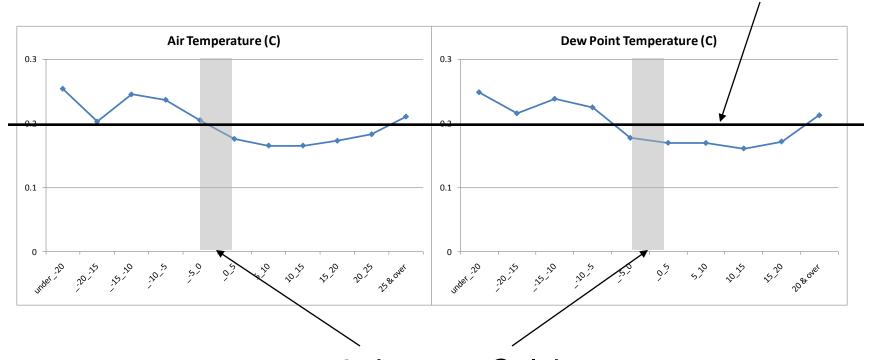
- The regression model implies linear effects, but...
  - Temperature changes may have greater effects around freezing
  - What is the critical visibility level?
  - Road temperatures are critical around freezing
  - What about correlations between some of the variables?
- Back to the raw data
  - Where are the tipping points above or below which the regression modeling may be effective?



## **Tipping Points**

 About 20% of the hours observed around the 4 stations had a crash

Average Crash Rate

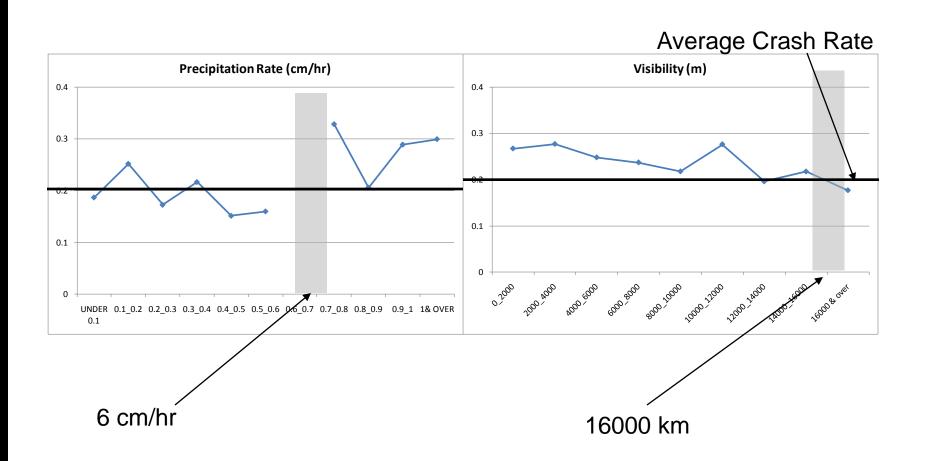


0 degrees Celsius



### **Tipping Points**

Precipitation Rate and Visibility

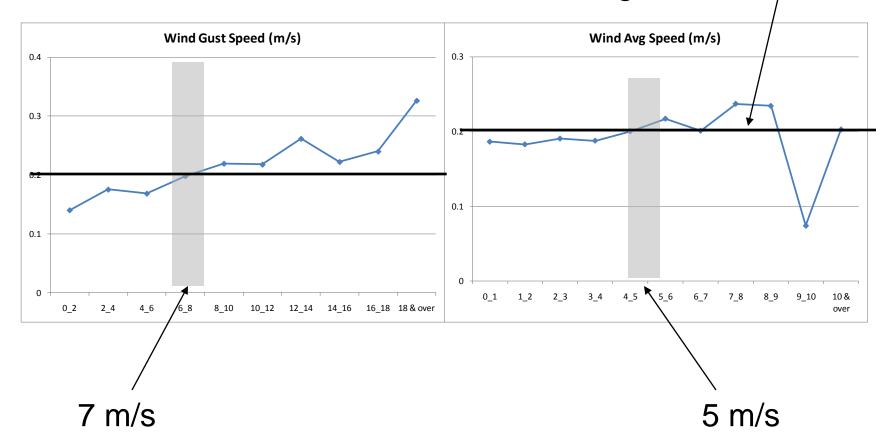




### **Tipping Points**

Wind Speed (average and gust)

Average Crash Rate



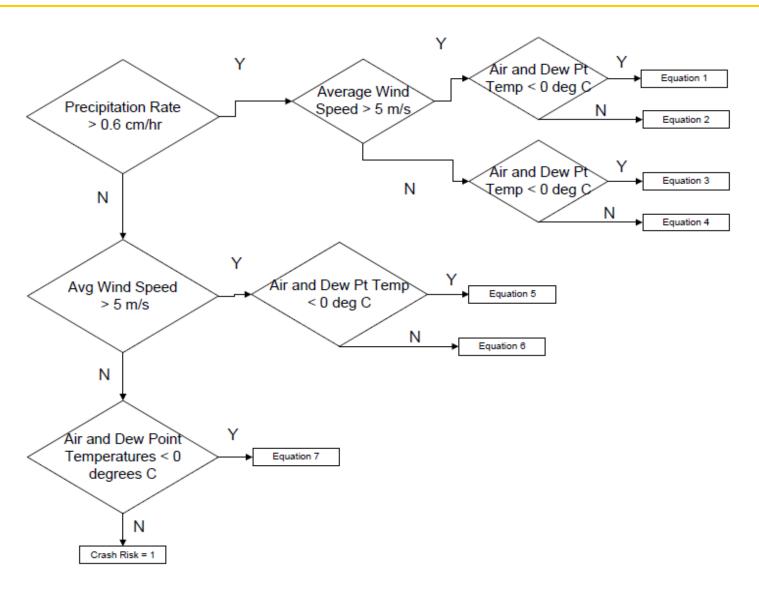


#### **Decision Tree Construction**

- A set of regression models applied under specific conditions.
  - Allows for evaluating continuous variables for regions of interest
- Evaluated subsets of data where crash risk was greater than 20% for all levels of other variables shown to be significant
  - i.e. the effect of dew pt, visibility, wind speed when air temperature is < 0 deg C.</li>



## **Crash Risk Algorithm**





#### **Equations**

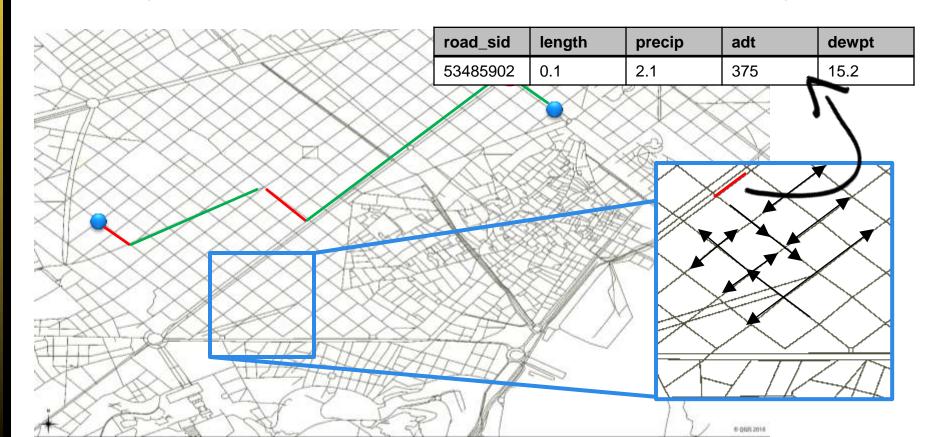
- For each path on the tree, a regression model was created as done originally.
- The exponential of the parameter estimate multiplied by the variable value yields the odds of a crash

```
CrashRisk_{Eq1} = e^{(0.6025+0.1716+0.2189)}
CrashRisk_{Ea2} = e^{(0.6025+0.1716)}
CrashRisk_{Ea3} = e^{(0.6025+0.2189-1.6789(AirTemperature)+1.4417(DewPtTemperature))}
CrashRisk_{Eq4} = e^{(0.6025)}
CrashRisk_{Ea5} = e^{(0.1716+0.2189)}
CrashRisk_{Eq6} = e^{(0.1716 - 0.0245(DewPtTemperature))}
CrashRisk_{Ea7} = e^{(0.2189 + 0.0130(AirTemperature) + 0.0438(AverageWindSpeed))}
```



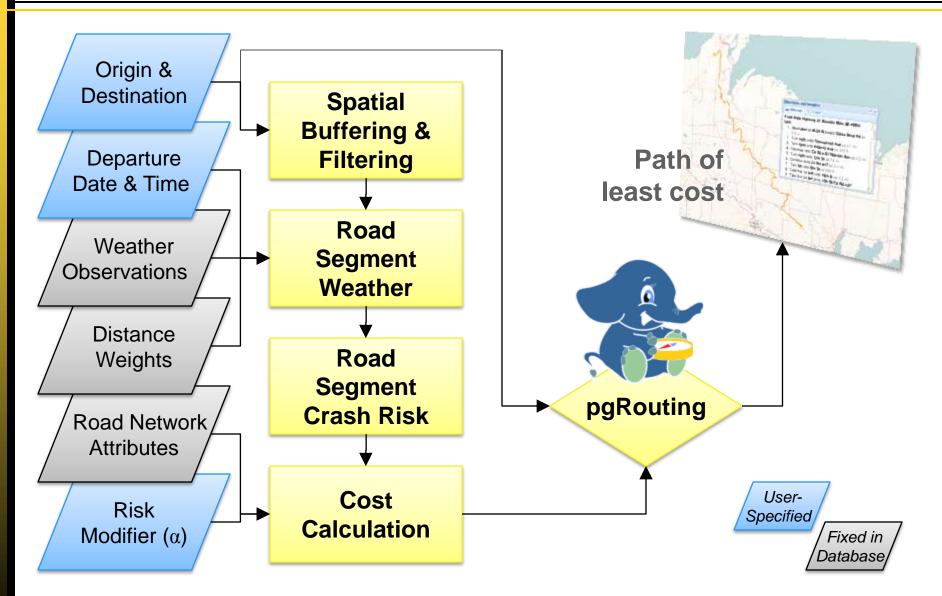
#### **Road Network Data**

- OpenStreetMap (OSM) data were loaded into a database to comprise the road network
- Length or travel time the typical cost of a road segment





## **Crash Risk Aversion Algorithm**





## **Road Segment Weather**

 Interpolate weather data for the road network using inverse distance weighting (IDW)

$$z(r) = \sum_{i=1}^{N} \left( \frac{w_i(s)z_i}{\xi} \right) \qquad \xi = \sum_{i=1}^{N} w_i \qquad w_i(s) = \frac{1}{distance(r, s)^p}$$

```
z_i Weather observation at a given CLARUS station
```

 $w_i$  Weight applied to the weather observation

r Road segment centroid

s Location of CLARUS station

*ξ* Normalization factor

p Power parameter (fixed at 2 in this application)

- IDW not the most rigorous spatial interpolation method, but best choice with only 4 CLARUS stations
- Inverse distance weights, calculated from road segment centroid, stored in the database for each road segment



 $\alpha$ 

#### **Crash Risk & Cost Calculation**

 Classical shortest time problem, but with crash risk considered as part of the cost

$$f(p) = cost_{p,t} = \alpha * traveltime_p + (1 - \alpha) * crashrisk_{p,t}$$

 $cost_{p,t}$  Cost of traversing edge p at time t

 $traveltime_p$  Time required to traverse edge p

 $crashrisk_{p,t}$  Crash risk associated with traversing edge p at time t

Weighting factor between 0 and 1; shortest path and least crash risk

$$crashrisk_{p,t} = \frac{\displaystyle\sum_{s \in S} \lambda_{s} crashrisk_{s,t}}{\displaystyle\sum_{s \in S} \lambda_{s}}$$

Crash risk for each nearby station by inverse distance weighting; in our problem, all four stations considered



# pgRouting

pgRouting

#### **PostgreSQL**

Adds types to typical data stored by Postgres

Operates on spatial representations enabled by PostGIS



**PostGIS** 

Find the least-cost path between edges



# **Routing Web Service**

- Apache server programmed in python with the django framework (and RESTful and AJAX-compliant)
- Client application written in Javascript using GeoExt (ExtJS); web mapping powered by OpenLayers
- Routing data sent in Javascript Object Notation (JSON)

