

CLARUS Road-Weather Routing for Crash Risk Aversion

John D Hill, Dept of Mechanical Engineering Michigan Technological University

K. Arthur Endsley, Michigan Tech Research Institute

Road Weather Management Stakeholder Meeting, Albuquerque, NM September 7, 2011

CLARUS Monitoring Stations

- 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?

- 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

Tipping Points

Precipitation Rate and Visibility

Tipping Points

- 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.

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

 $CrashRisk_{Eq1} = e^{(0.6025+0.1716+0.2189)}$ $=$ $CrashRisk_{Eq2}=e^{(0.6025+0.1716)}$ $=e^{(0.6025+)}$ (0.6025+0.1716)
(0.6025+0.2189–1.6789(AirTemperature)+1.4417(DewPtTemperature)) 3 *CrashRisk*_{Eq}₂ = $e^{(0.6025+0.1716)}$
*CrashRisk*_{Eq}₃ = $e^{(0.6025+0.2189-1.6789(AirTemperature)+1.4417(DewPtTemperature)}$ = $e^{(0.6025+0.1716)}$
= $e^{(0.6025+0.2189-1.6789(AirTemperature)+1.4417(L))}$ $CrashRisk_{Eq4} = e^{(0.6025)}$ $CrashRisk_{Eq5} = e^{(0.1716+0.2189)}$ $=e^{(0.1716+)}$ $(0.1716 - 0.0245(DewPtTemperature))$ $CrashRisk_{Eq5} = e^{(0.1716-0.0245(DewPtTemperature))}$ $=e^{(0.1716-1.05)}$ (0.1716–0.0245(*DewPtTemperature*))
(0.2189+0.0130(AirTemperature)+0.0438(AverageWindSpeed)) 7 $CrashRisk_{Eq6} = e^{(0.1716-0.0245(DewPtTemperature))}$
 $CrashRisk_{Eq7} = e^{(0.2189+0.0130(AirTemperature)+0.0438(AverageWindows) + 0.0438)}$ = $e^{(0.1716-0.0245(DewPtTemperature))}$
= $e^{(0.2189+0.0130(AirTemperature)+0.0438(A))}$

- 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

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

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

- *zⁱ* 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

 α

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*

p traveltime Time required to traverse edge *p*

p t crashrisk , Crash risk associated with traversing edge *p* at time *t*

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

$$
crash risk_{p,t} = \frac{\sum_{s \in S} \lambda_s crash risk_{s,t}}{\sum_{s \in S} \lambda_s}
$$

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

pgRouting

Routing Web Service

- Apache server programmed in puthon 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)

×.