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Impact of Climate Change on Influenza Mortality in the US

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08 January 2015



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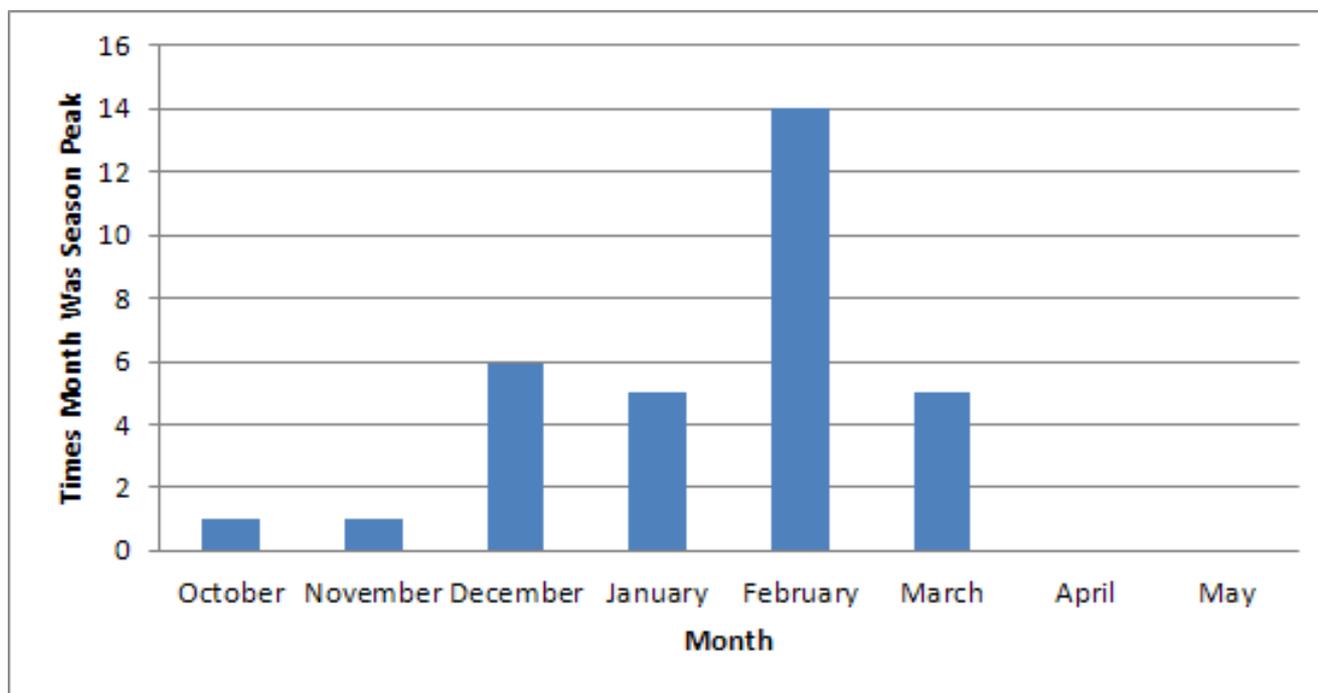
Introduction

- Each year approximately **5-20%** of US residents get the flu
- More than **200,000 people** are hospitalized
 - Usually contagious for 1-2 weeks
- Between 2008 and 2010, influenza related medical costs were estimated at over \$10 billion
- Relationship between temperature, humidity and influenza has been well studied in epidemiology
- Weekly data from 122 cities during 1970 – 2010 in US
- Non-parametric approach
 - Generalized Additive Model



Flu Season in the US

- Typical flu season in the US – October to March
- Flu activity peaks between December and February



Literature

- Positive association with humidity in El Salvador and Panama but negative association in Guatemala (Soebiyanto et. al, 2014 - PLOS ONE)
- Number of studies (Barecca and Shimshack, 2012; Deschenes and Moretti, 2007 and Martens, 1998) state that colder temperatures have greater influence on mortality
 - But high temperatures are may affect the inter-temporal distribution of mortality
- Based on laboratory experiments on guinea pigs, Lowen et al. (2007) show that both low temperatures and low humidity enhance viral stability





Climatic Variables and Influenza Interaction

- Low temperatures and/or extreme humidity increases mortality risk
- Impacts on cardiovascular and respiratory systems
- Low temperatures
 - Reduce blood flow and inhalation of cold air which may increase susceptibility
 - Limits exposure to vitamin D and increases indoor crowding





Climatic Variables and Influenza Interaction

- Low humidity
 - Leads to dehydration and increases spread of influenza by increased viral shedding
 - Also increases survival times of viral aerosols
 - It may also be connected through changes in the virus stability and transmission
- High humidity
 - Impairs body's ability to sweat and cool itself





Gap in the Literature

- Despite the hypothesized mechanisms - impact of climatic variables on influenza mortality are not well established
 - Especially on human population
 - Little empirical evidence
- Existing literature have used simplistic linear models
- Papers using more complex methodologies have focused on specific cities/regions
- Use of relative humidity
 - Strong positive correlation with temperature
- Assuming that patients are contagious for months
- Furthermore, understanding is limited to *a priori* assumptions of the pre-determined knots of the exposure variables' distributions





Gap in the Literature

- Large dataset at the city by week level reduces misclassification errors that may plague data observed on larger geographic scales like such as or nations
- We consider temperature and humidity simultaneously
- Use of GAM



Data

- Global Land Data Assimilation System (GLDAS 2)
 - Utilizes ground and satellite measurements
 - Models global terrestrial geophysical parameters
 - 0.25° by 0.25° spatial and 3-hourly temporal resolution
- Surface air temperature (K) and specific humidity
 - Specific humidity is the ratio between mass of water vapor and the mass of air (g/kg)
- Influenza mortality data from the Morbidity and Mortality Weekly Report (MMWR) on 122 cities
 - Weekly epidemiological digest for US published by CDC
- To obtain weekly data, we averaged the pixels and aggregated the 3-hourly data into daily data, and finally
 - Computed weekly maximum, minimum, and mean





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Descriptive Statistics

	Mortality	Temp-Max	Temp-Min	Temp-Mean	Humid-Max	Humid-Min	Humid-Mean
Median	3.00	22.93	7.24	14.06	0.011	0.005	0.007
Mean	5.08	21.03	6.79	13.01	0.012	0.006	0.008
Std. Dev	5.64	12.15	9.04	10.21	0.006	0.004	0.005



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Methodology: Generalized Additive Models

- No *a priori* reason for choosing a particular response function
- Allows response functions to be generated from the data
- GAM uses a link function to establish a relationship between the mean of the response variable and a smoothed function of the explanatory variables
- The general form is:

$$y = \alpha + \varphi + \sum_{i=1}^n f_i(X_i)$$

- The usual linear function of a covariate, $\beta_i \cdot X_i$, is replaced with f_i - an unspecified smooth function
- Non-parametric nature means that it doesn't assume a rigid form for the dependence of y on the predictors





Methodology: Generalized Additive Models

- Replaces the parameter “values times the predictor values” with a cubic spline smoother for each predictor
- Natural cubic spline smoothing functions of exposure variables and income were used
 - Removes the small variation while maintaining the major trend of each variable
 - Aim is to increase the efficiency in estimating the model
- Strength of GAM is its ability to deal with highly non-linear and non-monotonic relationships
 - GAM assumes additivity between predictors but allows for local nonlinearity in each predictor



Construction of Humidity Smoothing Terms

- Since only extremes matter - maximum and minimum humidity are transformed as:

$$H_{TR.Max} = H_{Max}/H_{Mean}$$

$$H_{TR.Min} = H_{Min}/H_{Mean}$$

- Estimates the effect of the extremes as multipliers of the average effect
- The transformed smooth splines are bounded on $(-\infty, 1)$ and $(1, \infty)$
- Separating out the conditional impact of the multiplier at any point on the mean spline





Methodology: Variables

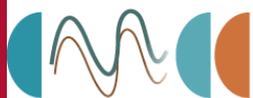
- γ_{iwy} is the natural logarithm of weekly per capita mortality in city i
- T and H are weekly temperature and humidity in city i
- I is the natural logarithm of the real income per capita of the Metropolitan Statistical Area that city i belongs to
- μ is a set of unrestricted time fixed effects and φ is a set of unrestricted city fixed effects
 - Control for unobserved confounding factors such as vaccination campaigns or the virulence of influenza strains



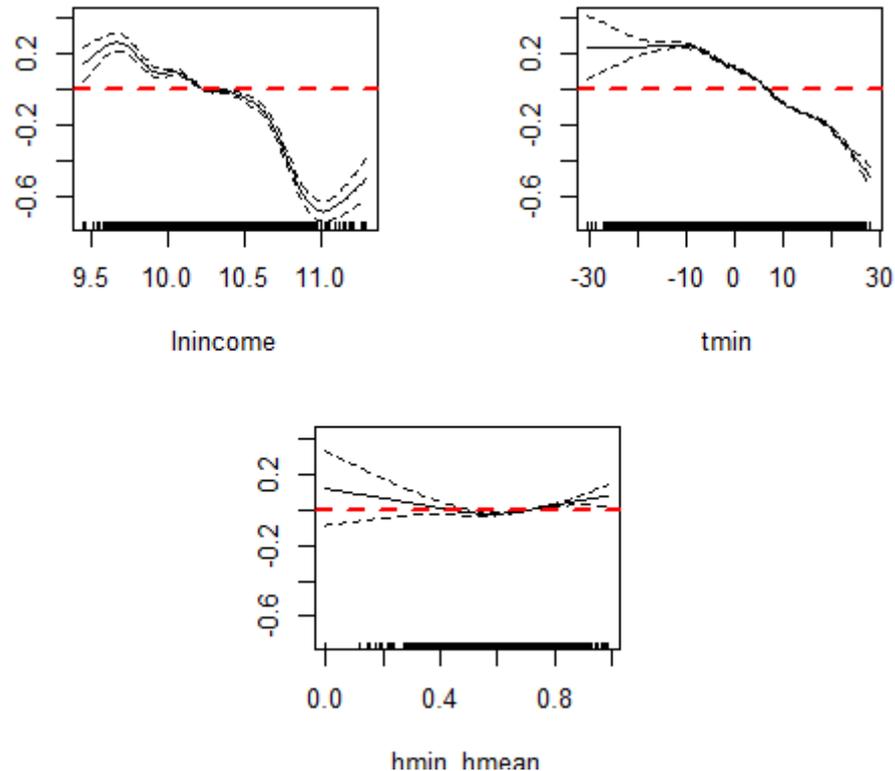


Results

- The variables are stationary
 - Im, Pesaran and Shin test
- We use Maximum Likelihood (ML) estimation method as its covariance parameters is more stable
- BIC test shows Min Temp and Transformed Min Humid regression to be most robust
- All the smooth terms are statistically significant
- The chances of influenza related mortality is higher at low income cities



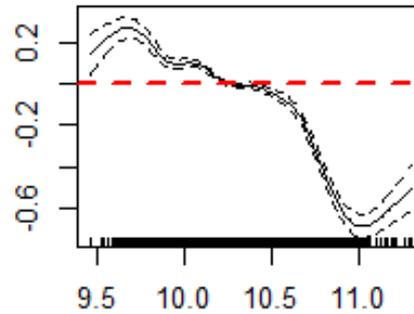
Results: Minimum Temperature and Humidity



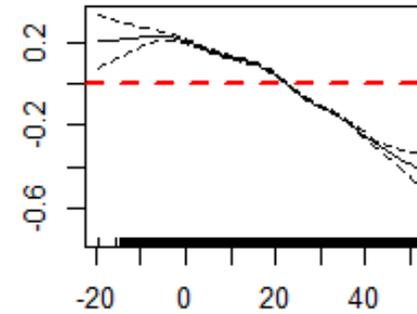
- Non-linear effect is evident
- More influenza mortality occurs at low temperature
 - Supports enhanced shedding theory at temperature of around -10°C
- Both very low and very high humidity are significant - *U shaped*
- BIC Test - This is the most robust specification



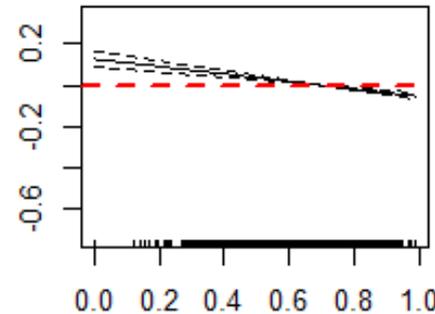
Results: Max Temperature and Min Humidity



Inincome



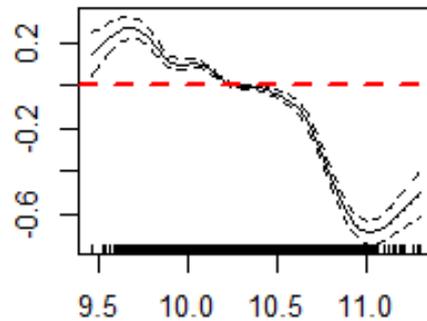
tmax



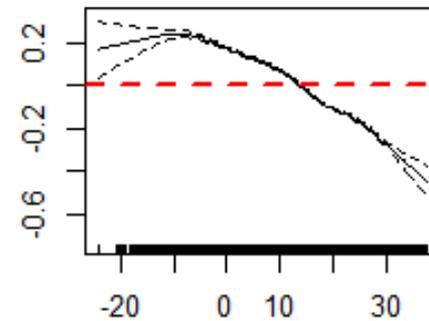
hmin_hmean

- Probability of influenza mortality is higher at low temperature and humidity levels

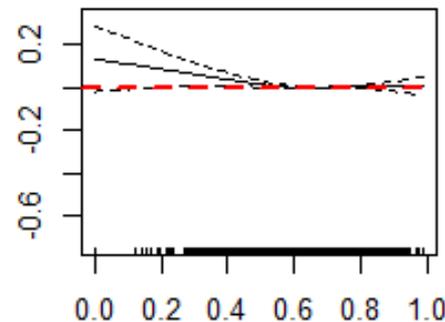
Results: Mean Temperature and Min Humidity



lnincome



tmean

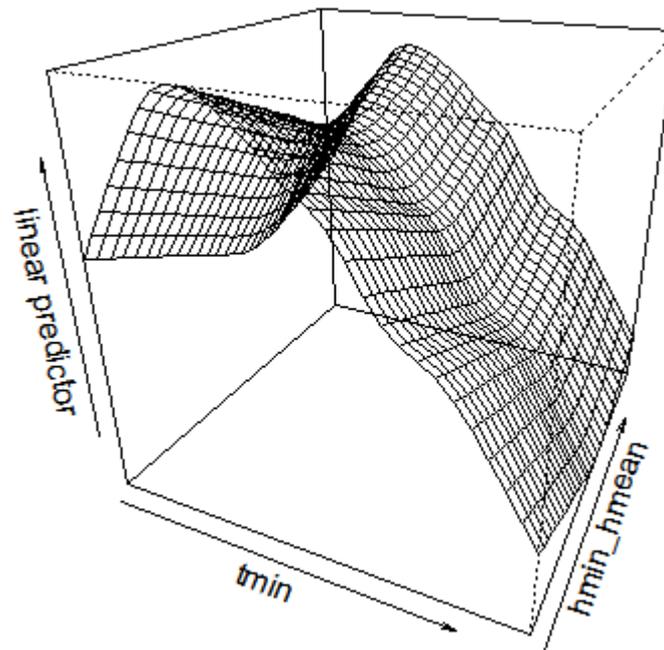


hmin_hmean

- Non-linear effect – specifications are robust
 - Do not change with mean, max, min

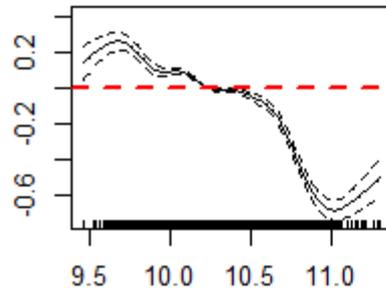


Results: Wireframe – Looking Beneath

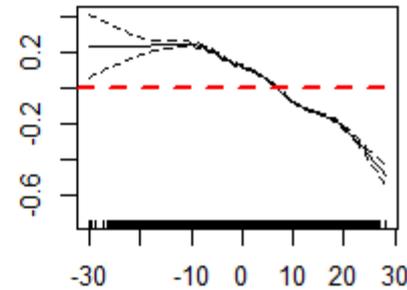


- Smoothing the marginal smooths of temperature and humidity
- Highest mortality risk at low temperature and low humidity

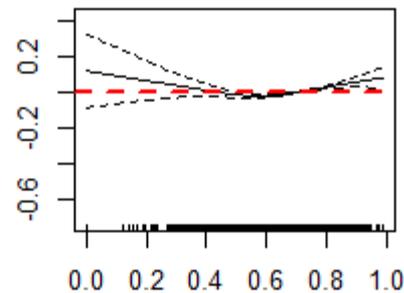
Results: Spatial Serial Dependence



lnincome



tmin



hmin_hmean

- One issue with GAM is serial dependence
- Assuming spatial serial dependence - we include
 - A bi-variate smoothing factor with latitude and longitude
- Response functions to not change





Discussion

- Different lag structures have no impact
- Data truncated for flu season do not change the shape of the response functions
- Neither does including a seasonal cyclical smooth term
- GAM results confirm the non-linear nature of the relationship





Conclusion

- Probability of influenza mortality is highest at the lowest temperature levels
 - Supports works under laboratory condition by Cannell et al., (2008) and Lowen et al., (2007)
- *U-shaped* specific humidity curve
 - Providing support for medical/epidemiological theory
- Influenza mortality declines as income rises
- Comprehensive evidence linking low T and extreme H to influenza mortality
- Next step is to use the results for projections
 - GCM data is being processed

