

epidemiology and disasters

response, recovery and research

Charles DiMaggio, PhD, MPH, PA-C

Professor of Surgery (Research) and Population Health
New York University School of Medicine
and
Director of Injury Research
NYU Bellevue Division of Trauma and Acute Care Surgery
New York, NY
Charles.DiMaggio@nyumc.org

24 April 2015

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

some conclusions about the role of epidemiology in disasters

- professional *duty* to be involved in disaster response
- public health *role* in disaster recovery and preparedness
- scientific *responsibility* to contribute to disaster research and risk assessment
 - disaster risk is complex
 - epidemiology can disentangle identify ways to mitigate and control risk
 - spatial analysis helpful in disaster epidemiology
 - Bayesian hierarchical modeling a great tool ...
 - ... but it comes with some additional analytic baggage

what is epidemiology?

the study of the distribution and determinants of disease in (human) populations

- basic science of public health - methods, study designs, statistical approaches
- descriptive - agent, host environments
- analytic - associations (causality?)
- statistics and chance

what is a disaster?

- quantitative**
- more than 100 deaths or \$1 million damage (Sheehan and Hewitt, 1969)
 - the bradford disaster scale
 - global annihilation = magnitude 10
 - 25K deaths = magnitude 4.3
- legal** “any natural catastrophe ...regardless of cause...which, in the determination of the President causes damage of sufficient severity and magnitude...”
- perception** “ordinary people will tell us what disasters are if we listen to them” (Kroll-Smith and Gunter, in Quarantelli, 1998)

an epidemiological definition of disasters

person the occurrence of more death, injury, disease, or psychiatric illness in a community than would be expected

time sudden, brief and intense causal mechanism

- natural vs. man-made
- “slow-moving disasters”

place

- circumscribed by geography or political boundaries
- ★ population and environmental vulnerabilities ★
“conditions prevalent in the...community (are) a better determinant of epidemiological impact than the physical characteristics of the event.” (Sapir and Lechat)

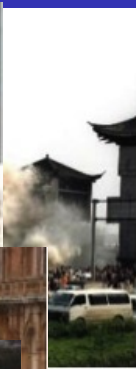














Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

Individual Disaster Response: 4 Groups of People

- killed
- injured - require immediate assistance / evacuation
- affected - can self evacuate
- involved - present at the scene (or nearby) willing and able to help
 - the *first*, first responders

Government Disaster Response: Expect Delays

- local
 - should be prepared to be self sufficient for the first 48 hours
- state
 - governor may declare “state of emergency” for state resources and support, mobilize national guard,
 - local military commanders may respond under “military support to civil authorities” doctrine for 72 hours
- federal
 - state request “national disaster” declaration (via FEMA) for federal resources
 - DMATs - disaster medical assistance teams, federally-sponsored, local groups medical professionals
 - NPS - national pharmaceutical stockpile
- in cases of terrorism, FBI is considered the “lead” agency

Incident Command Structure

- scalable (expands and contracts)
- flexible (any incident)
- fire, police, ems function essentially same as in non disaster setting
- incident commander (most senior fire or police, but may be you. . .)
and staff (public information, safety, medical, legal)
- operations (firefighting, security, EMS)
- planning (resources, documentation)
- logistics (communications, medical, food)

ICS Hierarchy

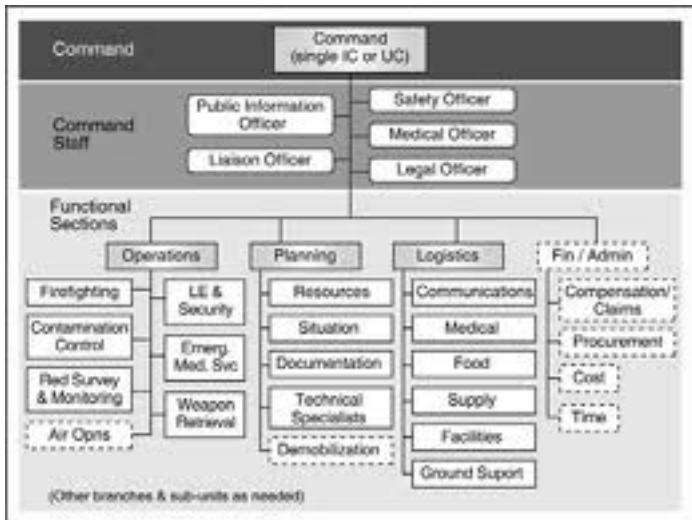


Figure: Incident Command Structure

Before a Disaster: Plan

- Where is the disaster plan?
 - multiple copies in all areas
- Who should I call?
 - fire, police, personnel, public health
- How should I call?
 - land line, cell, public safety radio, amateur radio, VOIP, satellite
- Should I secure my facility?
 - may need to lock down your facility to control traffic, separate access for potentially contaminated vs. non-contaminated (decon area), family access, “worried well”, media access
- Where can I get supplies?
 - plan on being on your own for first 48 hours

During a Disaster: Rescue

- Know Your Role
 - know where your disaster plan is
 - know who is in charge
- Risk from Patients
 - isolate for infectious diseases, biological agents (anthrax spores), radioactive or chemical agents (clothing, skin, breath, secretions)
- Risk from Environment
 - personal protective equipment (PPE); structural integrity (flooding, earthquake damage); contamination (water supply)
- Need for Evacuation
 - decontaminate for radiologic or chemical exposure *prior* to evacuation or transfer to ED

After a Disaster: Hazards

- communication is the first casualty
- trauma predominates early (first 2-3 days) then public health and chronic conditions
- electrocution kills
 - if water near electrical circuits, turn off power at main breaker
- people in vehicles are at greatest risk of drowning
- assume damaged buildings are unstable
- assume confined spaces are dangerous (toxic gases, explosions)
- dehydration and heat stroke in warm environments
- hypothermia in colder environments (working in water less than 75 degrees)
- avoid contact with potentially contaminated water, surfaces

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

three recent disasters

assessing the impact of 3 modern disasters

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - **three earlier disasters**
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

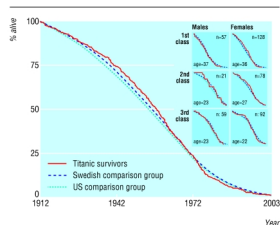
a tale of two disasters

parameter	'72 Managua	'71 California
richter scale	5.6	6.6
geographic extent	100km ²	1500km ²
population exposed	420K	7 million
dead	5,000	60
injured	20,000	2,540

their hearts will go on

case fatality rate

class	men	women	children	total
1	.67	.3	0	.38
2	.92	.14	0	.59
3	.84	.54	.66	.62
total	.82	.26	.48	.62



Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - **planning**
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

disaster phases

preparedness interdisaster training, building, warning

response emergency impact, extrication, treatment, food, shelter

recovery post-impact surveillance, rehabilitation, reconstruction

a haddon matrix for disasters

Noji and Sivertson

	PRE-DISASTER	DISASTER	POST-DISASTER
AGENT	identify potential terrorists		farming practices
HOST	discourage migration to disaster-prone urban areas	evacuation, shelter, (safe) food drops	prevent volunteer over-convergence
ENVIRONMENT	warning systems	levees	forensic engineering analyses

vulnerability assessment

- map risks
 - previous events by location and frequency and impact
 - population density and characteristics
 - environmental hazards and vulnerable structures
- inventory resources
 - infrastructure
 - equipment
 - personnel
 - transportation
 - ★ *communication*
- training
 - pod drills, table tops
 - sns, surge capacity
 - education, outreach

emergency aid

- most deaths occur during the first hour
- most life-saving aid is provided by families, friends and neighbors
- external aid is often too late, and frequently inappropriate

'external disaster relief should focus on reducing population vulnerability and invest in structural changes to health care'
(Sapir and Lechat)

health objectives

- prevent death
- treat injured and ill
- provide housing
- prevent illness due to communicable disease, lack of health care and malnutrition
- address acute psychiatric pathology and psychological distress

'generalized panic, paralyzing trauma or anti-social behavior rarely occur after major disasters' (Lazzari)

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - **surveillance**
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

observation and surveys

- data to make informed decisions
 - ‘...the sacrifice in promptness required to collect the information necessary to provide apt and well-directed aid is more than justified by the improved results’
- damaged infrastructure, no uniform definitions, multiple conflicting sources
 - ‘injury surveillance questionnaires ... prepared before a disaster (that) can be modified quickly’ (Noji)

surveillance

the ongoing systematic collection, analysis, interpretation and dissemination of health data

traditional in place prior to disaster, active vs. passive

non-traditional drop-in, syndromic (ED, pharmacy fills), relief workers, newspaper accounts, *spatial*

air and ground surveys



air and ground surveys



air and ground surveys



air and ground surveys



spatial analysis and mapping

person good methods to *analyze* nested, correlated data, e.g glm, mixed models

place good methods available to *identify* geographic clustering, e.g. SatScan, Moran's I

person-place Bayesian hierarchical models allow both identification and analysis

- *link...* mapping hurricane sandy damage
- *link...* images of japanese tsunami damage

Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

back to the future

dr. john snow (1813-1858)

“nature makes the experiments, and we watch and understand them if we can” (AB Hill)



back to the future

rev. thomas bayes (1702-1761)

“a Bayesian is one who, vaguely expecting a horse, and catching a glimpse of a donkey, strongly believes he has seen a mule”



Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

the problem(s) of spatial data

- *any* data analysis following a disaster is difficult
- spatial data has unique statistical issues
 - counts
 - irregularly arrayed (aerial vs. lattice)
 - population and geographic characteristics correlated in time and space
 - overdispersed (Poisson)

measuring risk spatially

- SMR

- y_i - observed count region i
- e_i - expected count region i
- θ_i - relative risk region i (unknown parameter)

- $smr = \frac{y_i}{e_i}$ - crude estimate for θ_i

- *but ... smr notoriously unstable and potentially misleading*

- non-independent observations, overdispersed, sensitive to denominator e_i
- WHO discourages use of smr's in maps

smoothing risk estimates

- number of approaches available
- Bayesian approach
 - demonstrated to be effective and stable
- establish the probability of a risk estimate (*the posterior distribution*) given the data we collected (*the likelihood*) and what we expected to see (*the prior distribution*)
 - $Pr[\theta|y] \propto Pr[\theta]Pr[y|\theta]$
- when in doubt, repeat: *the posterior equals the prior times the likelihood*

Gamma prior distribution

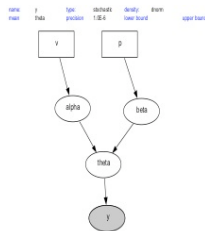
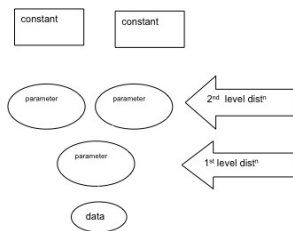
- the prior expectation of risk (θ) in spatial setting commonly defined as gamma ($\theta \sim \Gamma(a, b)$)
- $\mu = \frac{a}{b}$, $var = \frac{a}{b^2}$
- flexible
 - $\Gamma(1, b) \sim$ exponential
 - $\Gamma(\frac{1}{2}, \frac{1}{2}) \sim \chi^2$
- constrained to be positive
- null model: $\Gamma(a = b) \sim 1$

Poisson likelihood

- the distribution of rare events
- counts
- single parameter
 - $Pois(\lambda) = e^{-\lambda^k/k!}, \mu = \lambda, var = \lambda$
 - $Pr[k] = \frac{\lambda^k}{k!} e^{-\lambda}$
- observed spatial data: $y_i iid \sim Pois(\lambda_i = e_i \cdot \theta_i)$
- i.e. $E(y_i) = \lambda_i = \text{expected count} \times \text{risk}$
- *conjugate* to Gamma (same family)

hyperparameters

- Bayesian approach puts distribution on parameters
- so each parameter has its own paramters
- $y_i \sim \text{Pois}(e_i; \theta_i)$
- $\theta_i \sim \Gamma(\alpha, \beta)$
- $\alpha \sim \text{exp}(\nu)$,
 $\beta \sim \text{exp}(\rho)$



model statement

```
model{
  for (i in 1:m){
    y[i]~dpois(lambda[i])           # Poisson likelihood (data)
    lambda[i]<-e[i]*theta[i]
    theta[i]~dgamma(a,b)           # prior for relative risk
  }
  a~dexp(0.1)                       # hyperpriors
  b~dexp(0.1)
}
```

the log-linear transformation

- log link: $\ln(\lambda) = C1 + C2$
- where
 - $C1 = \beta x_i$ log-linear transformation
 - $C2 =$ other terms (e.g. predictors, random effect, spatial effect)
- create analytic models from gamma-poisson

the random effects term

- theoretically useful way to account for unstructured heterogeneity in data (spatial or otherwise)
 - variance ('noise') > accounted by model
 - ?unmeasured confounding
- practically
 - partition the error or residual term ($\hat{\epsilon} \sim nI(0, \sigma^2)$)
 - separate out variance component due to random effects, $\nu \sim nI(0, \sigma^2)$

conditional autoregression term (CAR)

- captures *structured* spatial variability
 - like a bookend to the *unstructured* random effects term ν
 - variation not accounted for by model or random effects term
- based on sets of 'spatial neighborhoods' (share a boundary)
 - $\mu_i \sim n(\bar{\mu}_{\delta_i}, \tau_{\mu} / n_{\delta_i})$
 - where δ is a neighborhood of the i^{th} region and n_{δ_i} is the number of neighbors
 - the mean θ or risk in a neighborhood is normally distributed, with $\bar{\mu}$ the average of the μ 's in the neighborhood, and σ equal to the σ of the neighborhood μ 's divided by the number (δ) of spatial shapes in the neighborhood

coding CAR in WinBUGS

- use the `car.normal()` distribution
- requires an adjacency vector, weights, and list of number of neighbors for each region

```
b[1:J] ~ car.normal(adj[], weights[], num[], tau.b)
```


the updated (spatial) gamma-Poisson model

log-link, random effects and CAR

- $y_i \sim \text{Pois}(e_i\theta_i)$
- $\ln(\theta) = \beta_i + \nu_i + \mu_i + \hat{\epsilon}$, where
- $\nu_i \sim \text{nl}(0, \tau_\nu)$
- $\mu_i \sim \text{nl}(\bar{\mu}_{\delta_i}, \tau_\mu/n_{\delta_i})$
- β_i is the intercept term and vector of regression coefficients for explanatory variables

spatial model statement

```
model{
for( i in 1 : m ) {
  y[i] ~ dpois(mu[i])
  mu[i] <- e[i] * rr[i]
  log(rr[i]) <- b0 +v[i] + h[i]   # h structured (spatial)
                                   # v unstructured(random)
  r[i]<-(y[i]-mu[i])             # r residual
  v[i]~dnorm(0,tau.v)           # normal prior on v
    }
  b0~dnorm(0,tau.b0)           # normal prior on intercept

  h[1:m]~car.normal(adj[], weights[], num[], tau.h) # car term
  tau.b0~dgamma(0.001,0.001)   # gamma hyperpriors
  tau.v~dgamma(0.001,0.001)
  tau.h~dgamma(0.001,0.001)
}
```

simulation in WinBUGS

- Monte Carlo Markov Chain (MCMC)
 - most reasonably realistic problems framed in Bayesian way do not have simple or closed solutions
- WinBUGS
 - implementation of MCMC
 - chooses samples using either Gibb's (for which named) or Metropolis Hastings
 - R2WinBUGS interface
- sample space and convergence (there is no free ride)
 - MCMC comes with additional responsibilities
 - is sampler moving accross the space?
 - did the chain converge?

non-convergence

- not sampling from the posterior distribution so results not valid
- more informative priors for problematic variables
- drop or add variables
- ★ re-parameterize to improve sampling efficiency
 - center
 - standardize
 - log transform

The BUGS Project - WinBUGS

http://www.mrc-biu.cam.ac.uk/bugs/winbugs/contents.shtml

The BUGS Project
winbugs

[Welcome Page](#)
[Latest News](#)
[Contact us/BUGS list](#)
[WinBUGS](#)
[New WinBUGS examples](#)
[FAQs](#)
[DIC](#)
[GeoBUGS](#)
[PKBUGS](#)
[Running from other software](#)
[BUGS resources online](#)
[WinBUGS development site](#)
[OpenBUGS site](#)

WinBUGS 1.4.x is now immortal!!!
 We no longer require users to register to obtain the key for unrestricted use...
 And the key is now everlasting!
[Click here to download the key](#)

Patch to upgrade to WinBUGS version 1.4.3 now available!
 (this patch is cumulative and contains minor fixes over 1.4.2)
[Click here for details](#)

Quick start

- Download and install WinBUGS 1.4.exe
- If installing on a 64-bit machine, you should download a [zipped version of the whole file structure](#) and unzip it into Program Files or wherever you want it.
- Download and install the [patch for 1.4.3](#)
- Get the [free key for unrestricted use](#) by clicking [here](#) - this can be used for multiple installations.
- See the [main BUGS page](#) for a summary of the different versions of BUGS available.

Contents

- Introduction to WinBUGS
 - Installing WinBUGS 1.4 in Windows
 - Obtaining the key for unrestricted use
 - Upgrading to version 1.4.3
- Problems reported in WinBUGS 1.4
- What else is available?

Introduction to WinBUGS

WinBUGS is part of the BUGS project, which aims to make practical MCMC methods available to applied statisticians. See the [main BUGS page](#) for a summary of the different versions of BUGS available.

WinBUGS can use either a standard 'point-and-click' windows interface for controlling the analysis, or can construct the model using a graphical interface called DoodleBUGS. WinBUGS is a stand-alone program, although it can be [called from other software](#). For a version that BUGS (BRugs) that sits within the R statistical package, see the [OpenBUGS site](#).

In the past, we have required that users register with us to obtain unrestricted use. In particular to give us some idea of the intended use of the program. This information has provided valuable feedback to ourselves and our funders. However, we no longer require users to register and the key for unrestricted use (of WinBUGS 1.4.x) can be downloaded from [here](#).

Watch [WinBUGS - The Movie!](#) for a short Flash illustration of the basic steps of running WinBUGS. (This movie features WinBUGS 1.3 - when you've watched it you will understand why WinBUGS 1.4 includes scripts to avoid all the clicking and pointing!).

All the documentation specific to WinBUGS is available on-line and is packaged with the program.

PDF versions of both the [WinBUGS 1.4 manual](#) (which DOES NOT include features in the 1.4.3 patch) and the [GeoBUGS manual](#) (which DOES feature the 1.4.3 patch) can now be downloaded, as can PDF versions of [Volume 1](#), [Volume 2](#), and [Volume 3](#) of the examples.

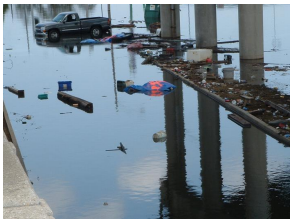
Obtaining the File(s)

Please read the [LICENSE AGREEMENT](#) for WinBUGS, before downloading anything.

This file uses a standard name WinBUGS1.4.exe for obvious reasons. The file will install a version how installation you to name the file with your number, and the key for unrestricted use, which is free.

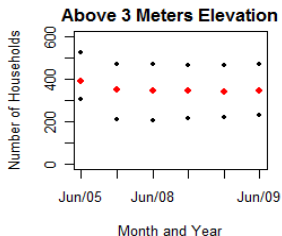
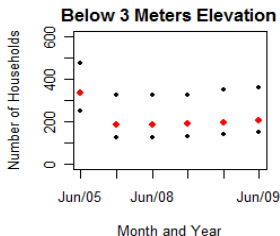
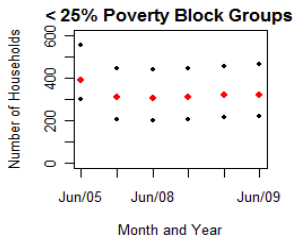
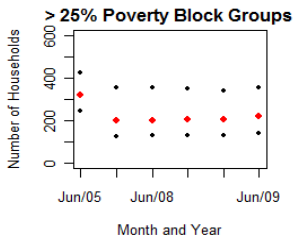
Outline

- 1 epidemiology and disaster response
 - incident command structure: the language of disasters
 - before, during and after a disaster
- 2 epidemiology and disaster recovery
 - three modern disasters
 - three earlier disasters
- 3 epidemiology and disaster preparedness
 - planning
 - surveillance
- 4 epidemiology and disaster research
 - Bayes, BUGS and R
 - orleans parish before and after hurricane katrina

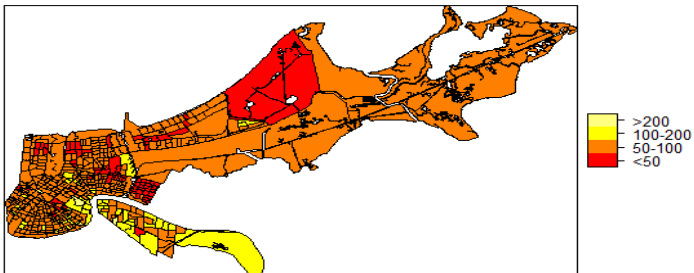


- repatriation measured by USPS delivery data
- June 2009 (observed) vs June 2005 (expected)
- θ ; estimates change in number of households receiving mail in a block group
- interested in *decreased* θ

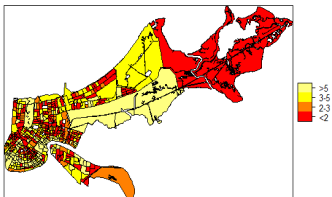
repatriation by SES and Geography



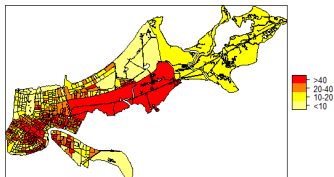
percent repatriation by block group



poverty and elevation choropleths



proportion below
poverty level



elevation above sea
level in meters

model statement

CAR, poverty, elevation, elev*pov

```

model{
  for(i in 1:m ){
    y[i] ~ dpois(mu[i])
    mu[i] <- e[i] * rr[i]
    log(rr[i]) <- b0 + b1*((elevation[i]-3.7)/3.6)
                  + b2*((poverty[i]-28.4)/18.2)
                  + b3*((elevation[i]-3.7)/3.6)*((poverty[i]-28.4)/18.2))
                  + h[i] # CAR term

    r[i]<-(y[i]-mu[i])          # r residual
    prob50[i]<-step(0.5-rr[i])  # step function exceedence
  }

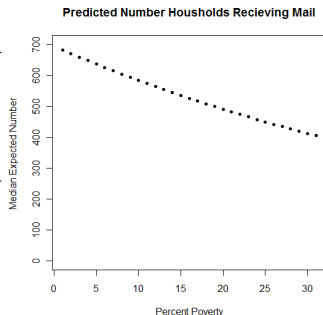
  h[1:m]~car.normal(adj[], weights[], num[], tau.h) # car term
  b0~dflat()          # priors
  b1~dnorm(0,0001)
  b2~dnorm(0,0001)
  b3~dnorm(0,0001)
  tau.h~dgamma(0.001,0.001)
}

```

models compared using DIC (Bayesian analog of AIC)

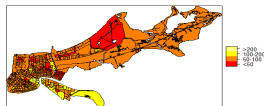
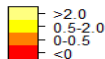
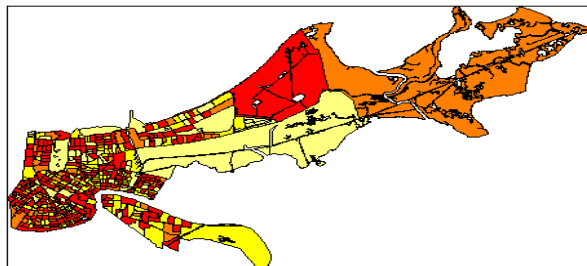
some numeric results

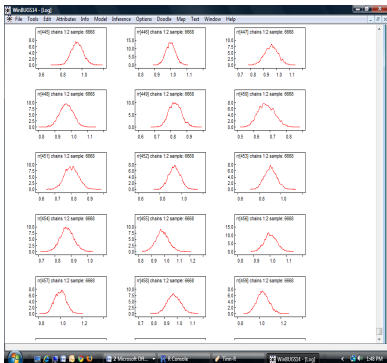
parameter	mean	credible interval
intercept	-0.36	(-0.37, -0.35)
elevation	0.029	(-0.009, 0.07)
poverty	-0.14	(-0.20, -0.10)
elev*pov	0.07	(0.04, 0.11)



smoothed repatriation estimates

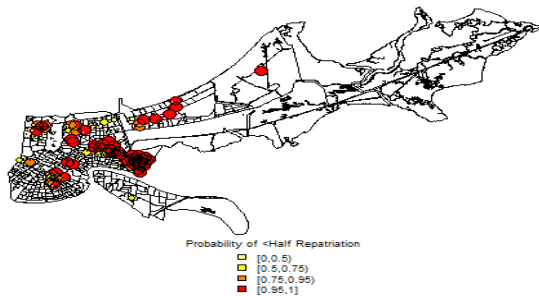
adjusted for poverty and elevation



posterior probability of θ_i 

can be used to
map
exceedence, or
probability
greater than x

exceedence map



conclusions (redux)

- as public health *professionals* epidemiologists have a *duty* to be aware of and involved in disaster response
- as a *public health* discipline, epidemiology has an essential *role* in preparedness and recovery
- as the basic *science* of public health, epidemiology has a fundamental *responsibility* to contribute to research that advances the body of knowledge about disasters
 - socioeconomics, behavior and geography intersect and interact in complex ways to mediate disaster risk
 - the role of epidemiology is to disentangle these competing and interacting risk to identify ways to mitigate and control the societal and health effects of disasters
 - spatial analysis can extend epidemiologic methods following disasters
 - Bayesian hierarchical modeling a great tool to explain spatial patterning of health outcomes and risk
 - but it comes with some additional analytic baggage (prior specification, convergence, interpretation)



Figure: Boston Sunrise