Using Federal Environmental Data for Exposure Assessment in Epidemiologic Studies of Cancer

Mary H. Ward, Ph.D.

<u>wardm@mail.nih.gov</u>

Occupational and Environmental Epidemiology Branch
Division of Cancer Epidemiology and Genetics



Geospatial Statistics, Tools, Data, Practices, Opportunities and Challenges in the Federal Agencies Geospatial Interest Group of the Federal Committee on Statistical Methodology

Linkages to environmental data – NCI studies of cancer etiology

- U.S. Census data (e.g. income, education, housing, population density)
- Land use data (e.g. roads/traffic, agricultural fields [pesticides])
- Environmental monitoring data:
 - Drinking water contaminants (public water supplies)
 - Modeled levels in private wells
 - Industrial emissions (Toxic Release Inventory, Dioxin Emission databases)
 - ❖ Air pollutants (e.g. PM_{2.5})
 - Ultraviolet light

Using Geographic Information Systems for exposure assessment

 Mapping the study population, exposure sources, and exposure assessment (fate and transport modeling)

Nuckols JR et al., Environ Health Perspect; 2004

- Issues/Examples:
 - Accuracy of locations (geocoding)
 - Agricultural pesticides and emissions from animal feeding operations
 - Residential mobility
 - Drinking water contaminants
 - Public water supply monitoring data
 - Private wells modeled estimates

Locate the study population (and exposure sources)

- Global Positioning System (GPS)
 - High accuracy (m)
 - Easily added to home visit
 - Field studies can track individual activity patterns (e.g. commuting)
 - Not always feasible especially for large cohort studies
 - Geocode current and past addresses

NCI-SEER NHL case-control study

- Aim: To identify potential environmental causes of NHL
- Study design:
 - 1321 cases (diagnosed 1998-2000), 1057 controls
 - SEER cancer registries: Detroit, Seattle, Los Angeles County, State of Iowa
 - ❖ Age at diagnosis: 20 74
 - * Home visit, questionnaire:
 - Residential History
 - House dust samples
 - Blood samples
- Exposure Period: lifetime history (1923 2000)

Study area and GPS locations of current homes in the NCI-SEER NHL case-control study



Is geocoding accurate enough?

 We compared geocoded addresses to GPS measurements at home interview

 Calculated the positional accuracy for residences located inside and outside town boundaries (rural) in lowa

Positional accuracy (m) by location of home (rural, in town)

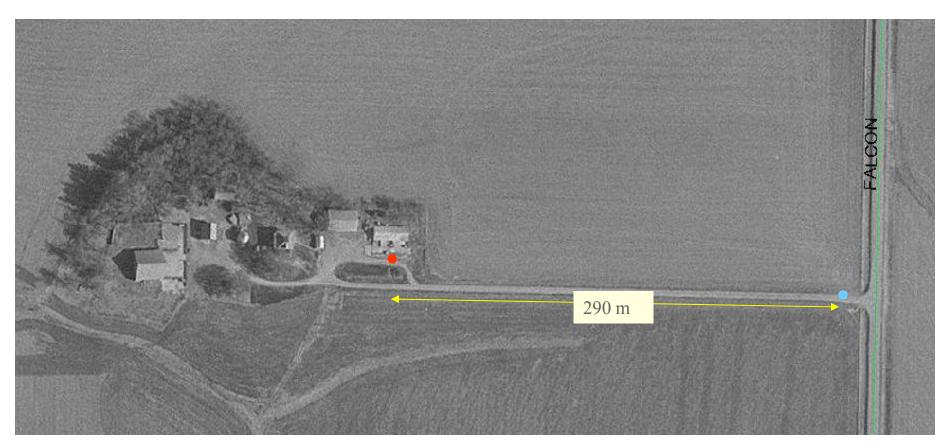
	Commercia	l firm	Geocoding in ArcGIS		
	Town (n=159)	Rural (n=56)	Town (n=152)	Rural (n=53)	
Median (IQR)	50 (28-83)	212 (89-747)	56 (36-92)	88 (44-254)	
p-value	<0.0001		0.0013		

- Rural addresses had greater errors
- Town size did not affect the positional accuracy for 'urban' addresses by either method

Agricultural Health Study

- Prospective cohort of pesticide applicators and their spouses in IA and NC in 1993-1996; follow-up interviews in 2000-3 & 2007-10
- ~90,000 farmer/pesticide applicators and spouses
- >150,000 addresses including past homes from short residential history
- ~66% of homes are rural
- Compared rooftop locations (digital orthophotos),
 E911 locations, and geocoded addresses

Distance between the E911 address location (blue) and the roof top location (red)



Orthophoto is 2 feet (0.6 m) resolution

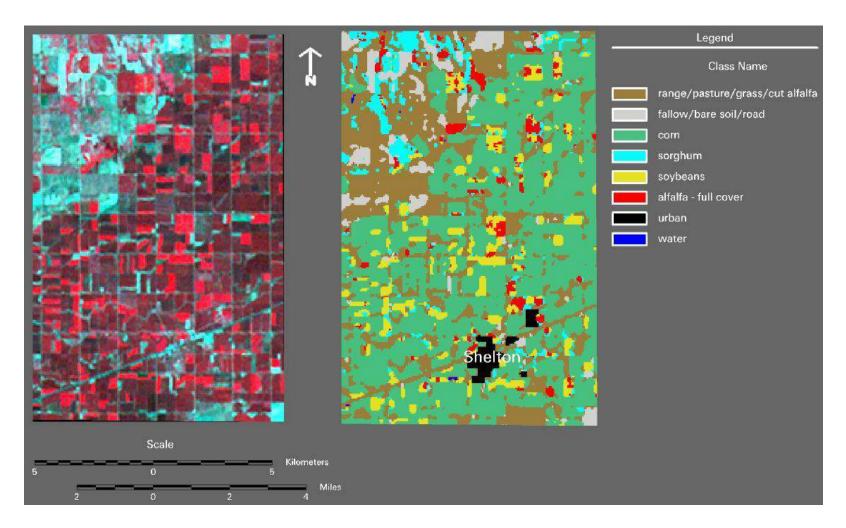
Positional error by geo-location method

Comparison	Median (meters)	Interquartile range (m)
Overall: Address-match vs. Rooftop	90	47-215
E911 vs. Rooftop	39	22-61
Rural: Address-match vs. Rooftop	147	78-353
E911 vs. Rooftop	51	39-83
In-town: Address-match vs. Rooftop	45	27-68
E911 vs. Rooftop	19	17-23

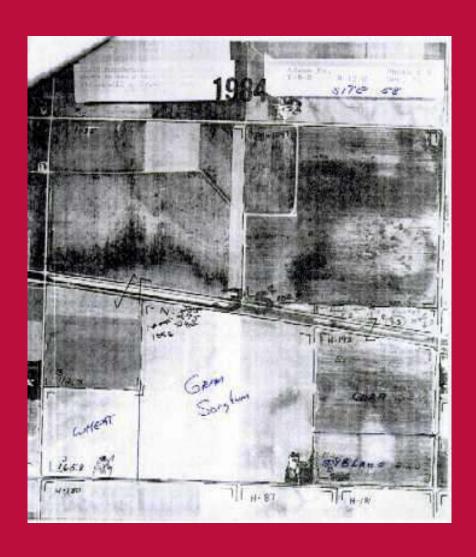
Estimating residential exposure to agricultural pesticides – NCI-SEER NHL study



Historical NASA Landsat imagery can be used to create crop maps

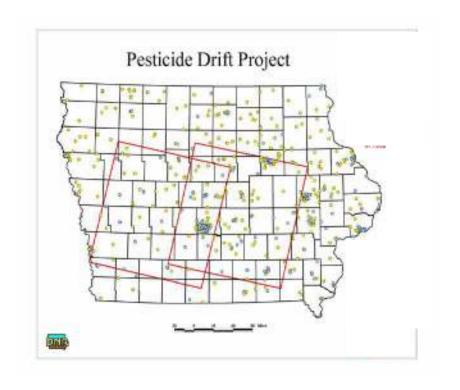


Farm Service Agency has historical aerial photographs with locations of crops



Estimating agricultural pesticides: NCI-SEER NHL case-control study

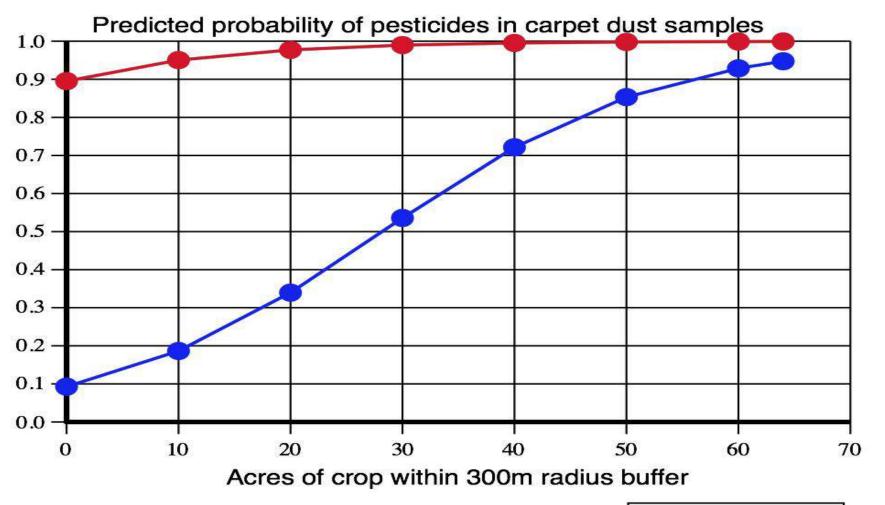
- Satellite imagery used for crop maps
 - **1998-2000**
- Dust samples
- 14 agricultural herbicides



lowa population with agricultural crops within various buffer distances of residence

Buffer (meters)	All residences	Within towns
100	29%	18%
250	44%	38%
500	60%	52%

Probability of detecting agricultural herbicides in homes of farmers (red) and nonfarmers (blue)

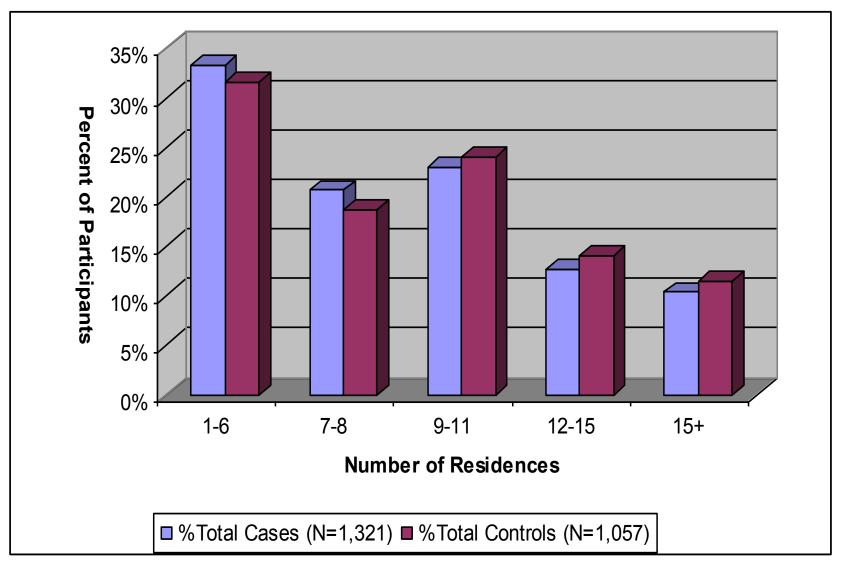




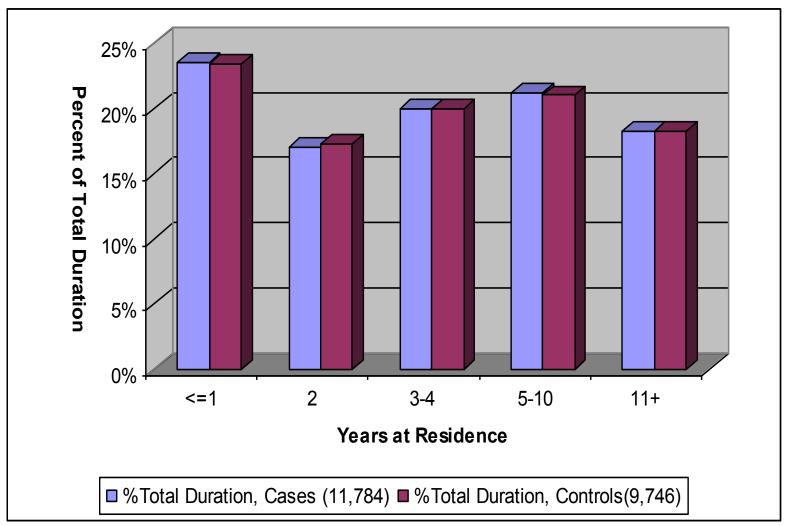
NCI-SEER NHL Study: Current and past homes



NHL Number of Residences



NHL Duration at each Residence

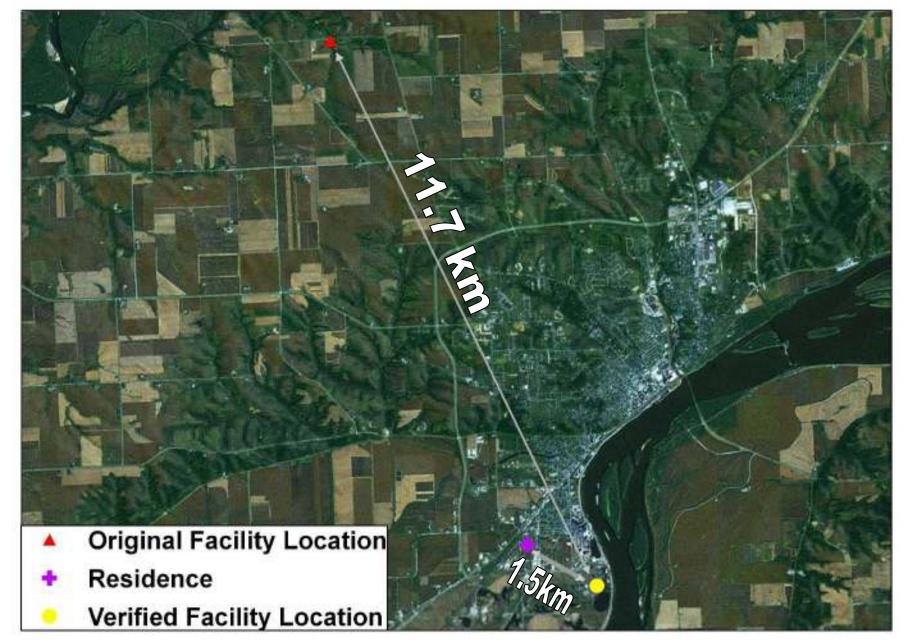


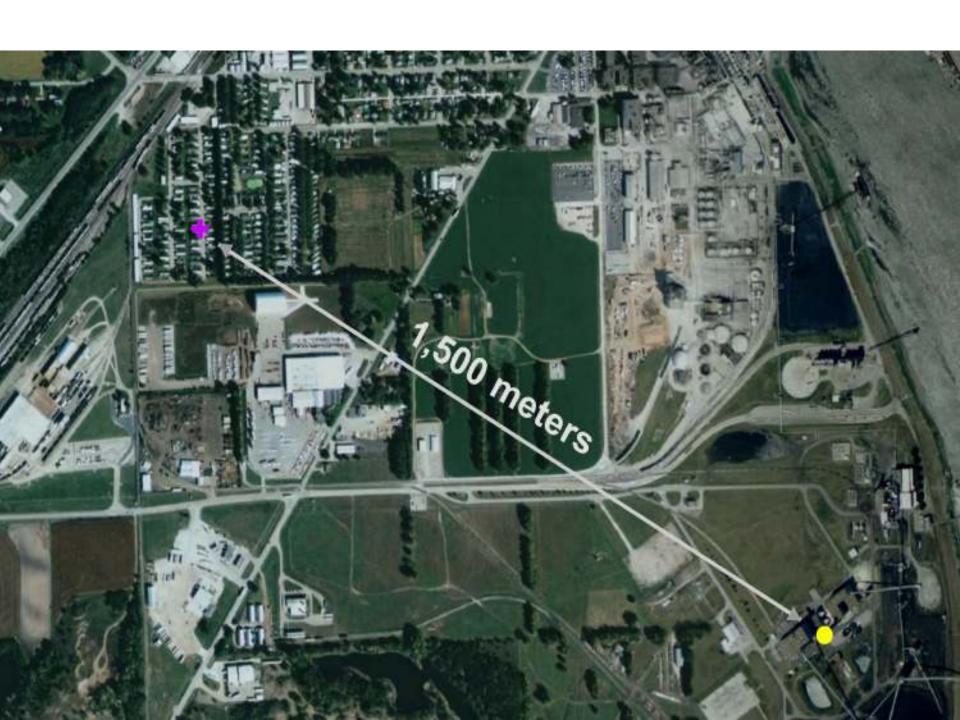
DIOXIN EMITTING FACILITIES EPA INVENTORY, 1987 and 1995



EPA: An inventory of sources and environmental releases of dioxin-like compounds in the United States for the years 1987, 1995, and 2000. U.S. Environmental Protection Agency; 2006.

Original and Verified Facility and Residence





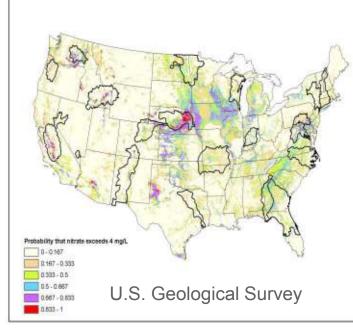
Distance between EPA Facility Geocode and Verified Location (Km)

	Cement Kilns	Coal Power Plants	Hazard Waste Incinerator	Medical Waste	Municip. Waste	Sewage Sludge incinerator
25th	1.2	0.05	4.4	0.0	3.2	1.0
Median	4.0	0.12	11.5	0.0	7.8	4.6
75th	6.7	0.45	32.8	2.7	29.0	23.4
Mean	7.0	0.93	23.5	48.7	23.1	13.8

Nitrate in drinking water: Sources and exposures

- Nitrogen fertilizers, animal and human waste
- Maximum contaminant level: 10 mg/L as NO₃-N
- Highest exposures: private well users
- Measurements are sparse





Drinking water nitrate and cancer in two cohorts in agricultural areas

- Iowa Women's Health Study (IWHS):
 - 42,000 postmenopausal women
 - Drinking water source and duration in 1989
 - 73% use public supplies, 25% private wells
 - Many surface water supplies
- Agricultural Health Study (AHS):
 - 60% use private wells

Iowa Women's Health Study: Public water supply exposure assessment

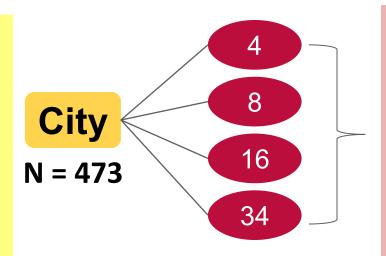
Median duration (years)



Public Water Supply Monitoring Data, 1955-1988

 NO_3-N

Total Trihalomethanes (THMs)



Exposure Metrics

Average NO₃-N and THMs

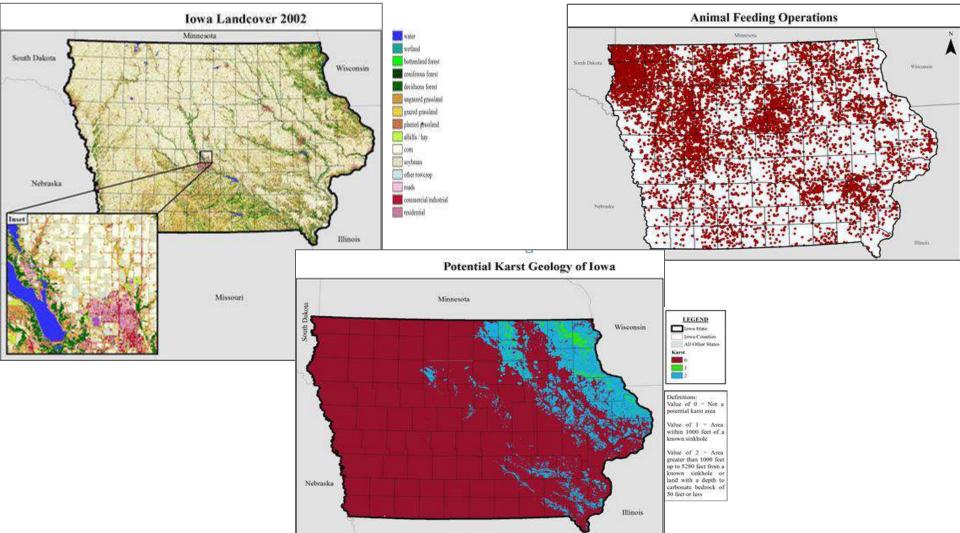
Years >1/2
Maximum
Contaminant Level

82% used supply 16+ years

Ward MH et al., *Epidemiology*; 2010 Inoue-Choi M et al., *Int J Cancer*; 2014 Jones RJ et al., *Environ Health Perspect*; in press

GIS-based model of nitrate in private wells

- ~34,000 nitrate measurements (1980-2000s)
- Evaluated >150 variables (e.g., land use, animal feeding operations, geology, soils)



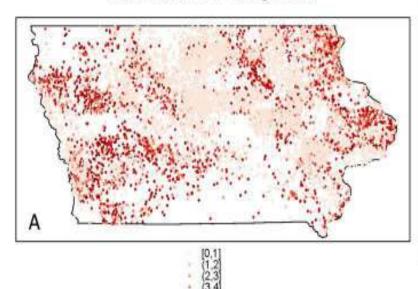
Random forest model performed best

- -66 variables explained 77% of variation in training dataset:
 - Well depth
 - Geologic features karst geology, sinkholes
 - Slope, elevation
 - Animal feeding operations
 - Agricultural land (1990)
 - Precipitation
 - Soil characteristics at well screen
 - Year

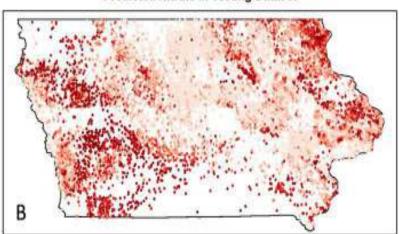
Sensitivity and Specificity (5 mg/L)

NO₃-N Observed <5 mg/L ≥5 mg/L Sensitivity 2615 2598 (67%) mg/L (14%)NO₃-N **Predicted Specificity** <5 1280 15660 (86%)(33%)mg/L

Observed Nitrate in Testing Dataset



Predicted Nitrate in Testing Dataset



Wheeler DC/Nolan BT et al, STOTEN; 2015

GIS for exposure assessment - summary

- Assess exposures not easily obtained by questionnaires or biomonitoring
- Need to assess and quantify uncertainty in spatial data
- Validation of GIS-based exposure metrics for assessing individual exposure
- Include activity patterns to refine exposure assessment

Nuckols JR et al, *Environ Health Perspect;* 2004 Ward MH and Wartenberg D, *Am J Epidemiol*; 2006

GIS for improving exposure assessment in cancer studies

- Will be increasingly useful in future data availability, improved technology for satellite imagery, geocoding
- Requires an interdisciplinary approach:
 - Geographers, environmental engineers, chemists, environmental scientists, epidemiologists

Nuckols JR et al, *Environ Health Perspect;* 2004 Ward MH and Wartenberg D, *Am J Epidemiol*; 2006

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NCI:

Joanne Colt

Maki Inoue-Choi

Rena Jones

Jay Lubin

Anjoeka Pronk

Westat:

Matt Airola

Abigail Flory

