

Visualization and Analytic Methods for the Tracking of Birth Outcomes and Traffic Exposures

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Overview

- Needs of Visualization for Tracking
- Mapping Disease Rates

 Failure of mapping in discrete areas
 Density estimation
 Strengths and Weaknesses

Modeling of Traffic-Exhaust Pollution
 Methods: Cost-Benefit Analysis

Needs of Data Visualization for Tracking

- We need a system that is continuous and ongoing
- We need web-based tools that allow public access and ability to interface with data
- We need to preserve data confidentiality and privacy
- The system is not complete until those who need information
 - Know the information exists
 - Know where to find it
 - Know what it's good for
 - Are able to access and interpret it

Mapping Disease Rates

Failure of mapping rates in discrete areas

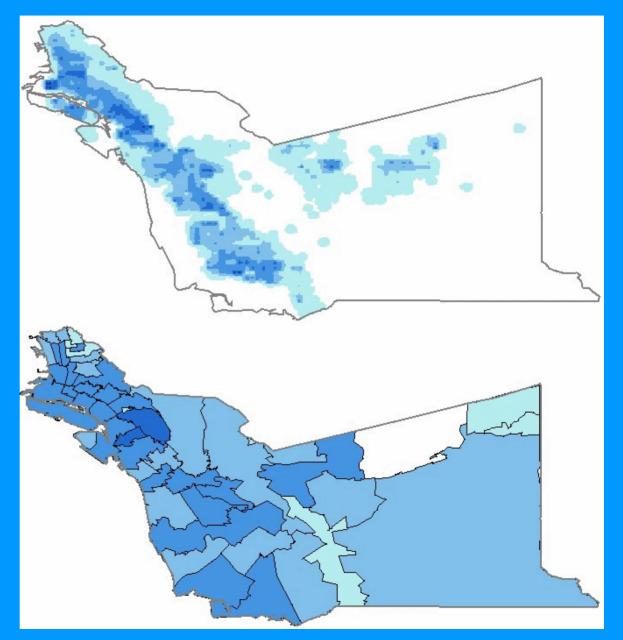
Sample size problem: law of small numbers
Instability of rates with small denominators
As areas get smaller, variability increases
Visualization fails
Political boundaries change over time (e.g. ZIP codes)

On-going monitoring and dissemination of information on the distribution of environmentally related disease

Pilot Project 1

Pre-term birth rate, Alameda County, 2001 (By density estimation and by zip)





Density Estimation: Strengths

Methods of Openshaw, Rushton

Produce continuous surface of rates

Preserves data confidentiality

More accurately reflect reality

Data Restrictions

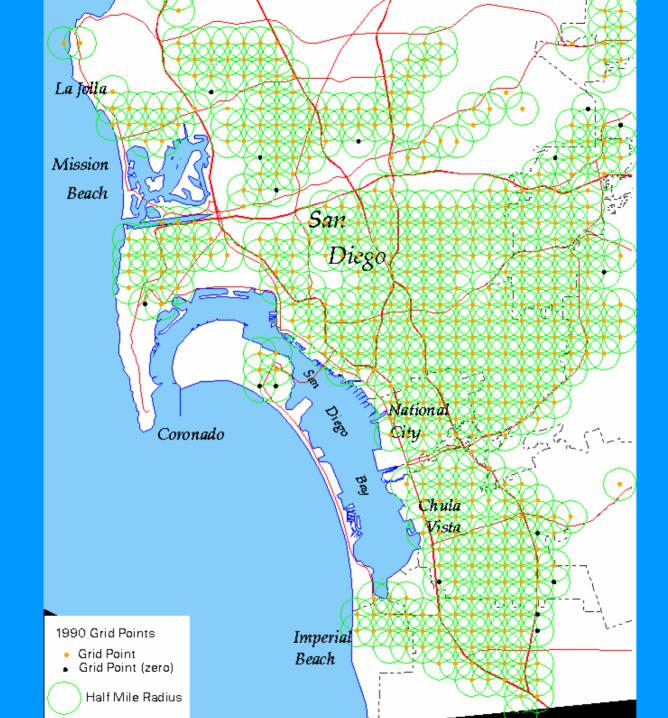
Restricted births to:

completed birthweight
geocoded address
downtown San Diego and nearby areas
compatible birthweight and gestational age

- 16,385 births 1980- 24,274 births 1990

Spatial Distribution Method

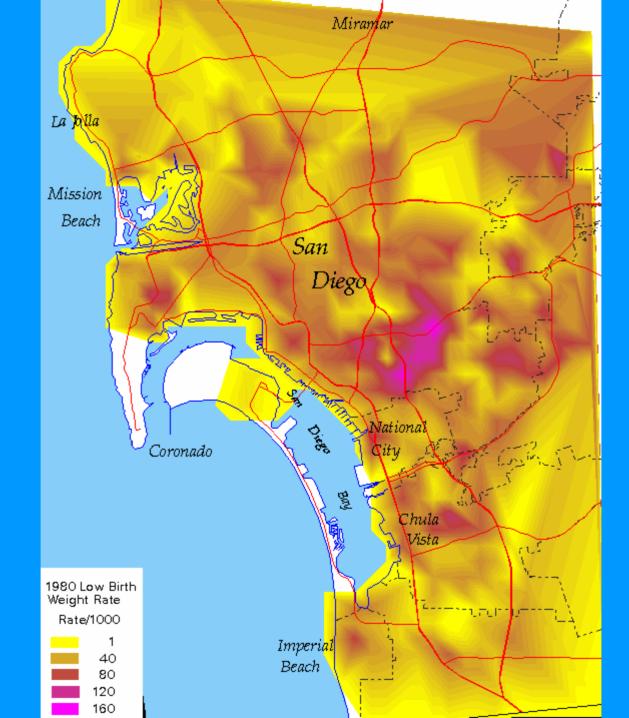
(1) Generated Uniform Grid 0.5 miles apart (spatial filters)

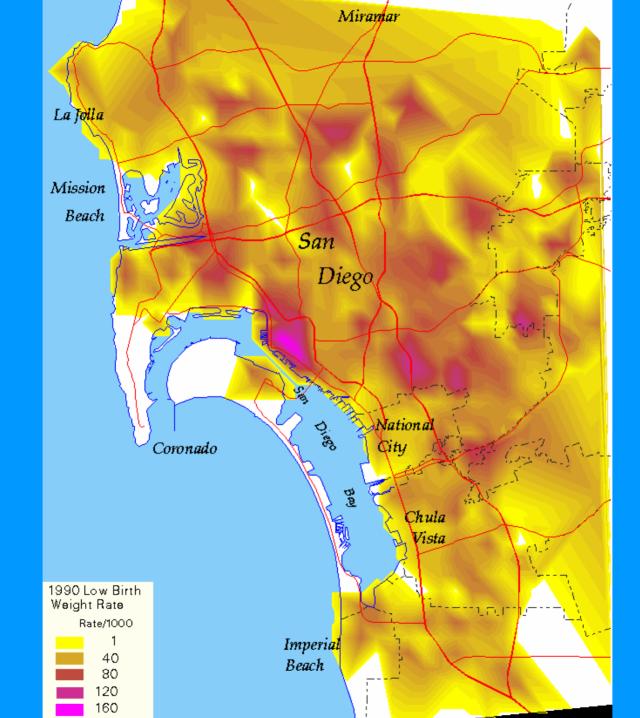


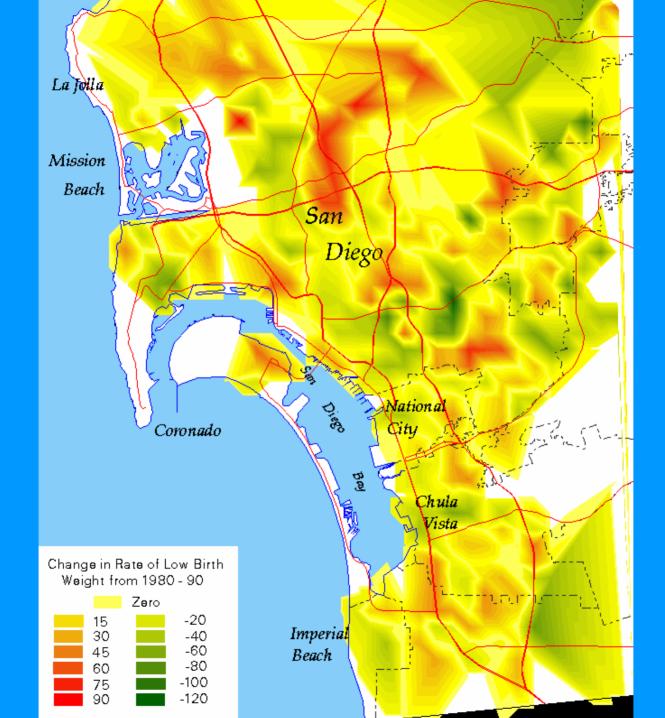
Spatial Distribution Method

(2) Identified all births within a 0.5 mile spatial filter (min. of 40 births to compute rate)

3) Compute LBW rates and made contour maps





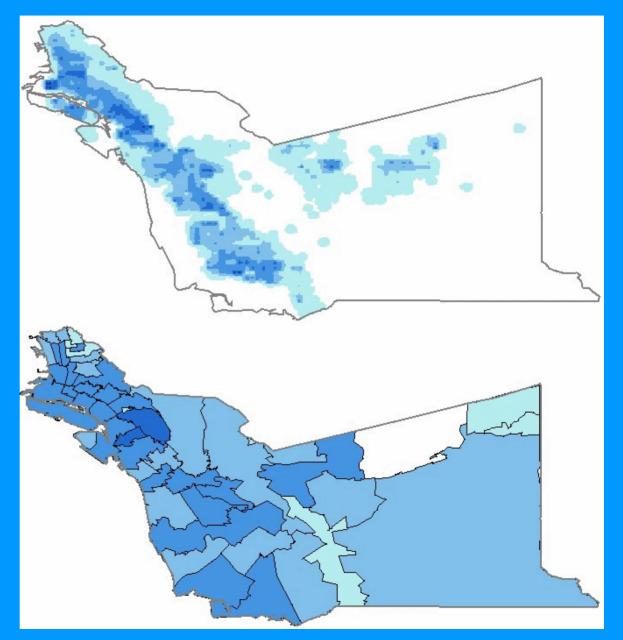


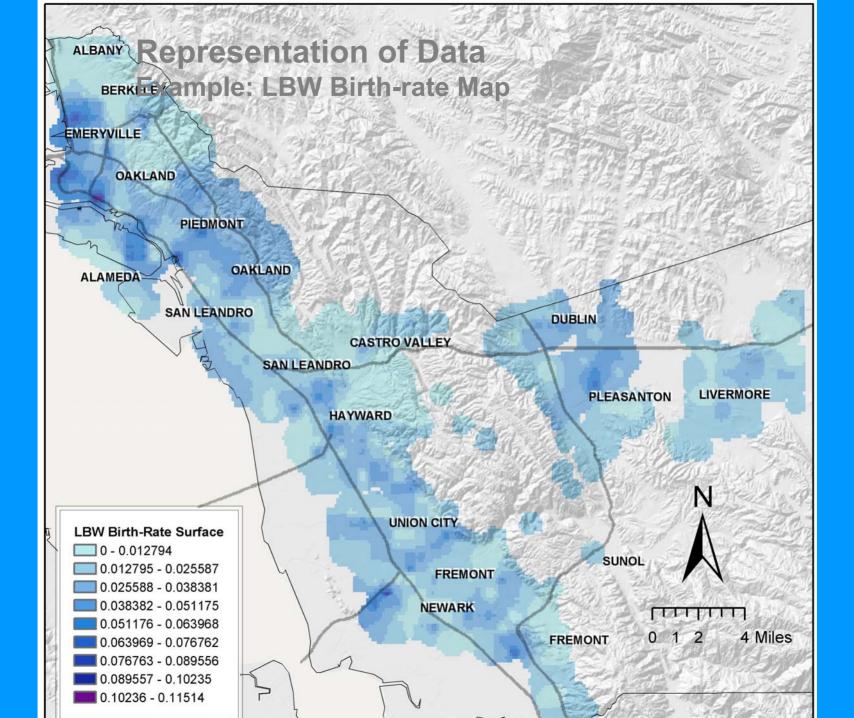
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Pilot Project 1

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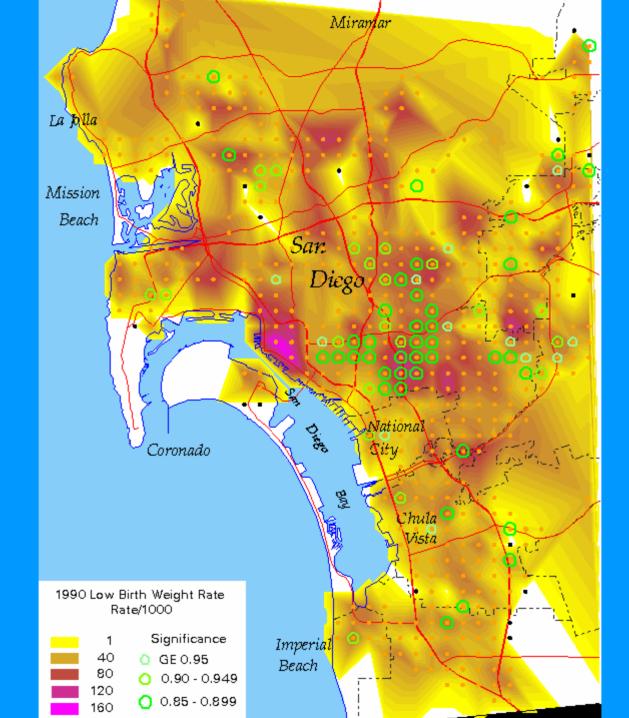
Statistical Significance (Monte Carlo Simulations)

- Assign each birth equal probability of being a LBW birth
- Assign a random number (1-1000) for each birth
- Classify each birth as LBW or non-LBW
- Compute LBW rate at each grid point 1000 times

Statistical Significance (Monte Carlo Simulations)

• Compute the % of simulated rates which are less than the observed rate.

Make map of statistical significance



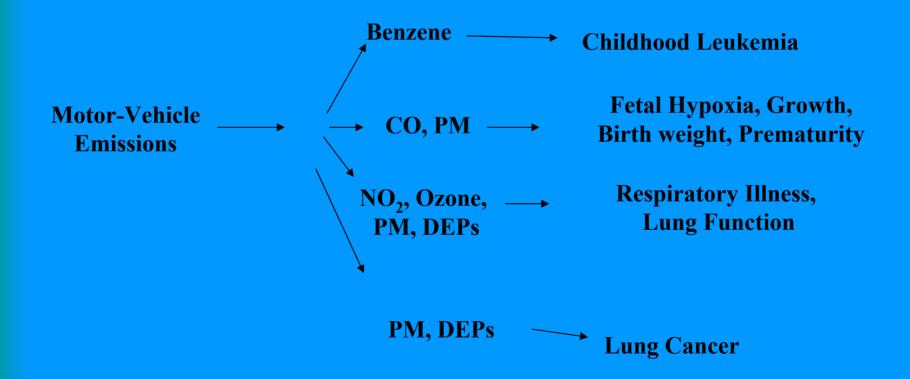
Density Estimation: Weaknesses

Need for geocoded denominator

Need for sensitivity analysis:
 – size of filter needs to change in relation of density of health events?

 Spatial autocorrelation an issue in analysis: Spatial regression

Health Effects Associated with Motor-Vehicle Emissions



Modeling of Traffic Exhaust Pollution

- Proximity Analyses
- Dispersion Models
- Land-Use Regression
- Integrated-Meterological Models

Simple to Complex Measures of Traffic-Related Pollutants

Complex '

More complex modeling (e.g. ADMS- Urban)

GIS w/ dispersion modeling GIS w/ regression-based models

Residence near fixed air monitors

Distance-weighted traffic volume

Simple

Neighborhood vehicle density

Limitations:

Computationally complex, cost

More data inputs (e.g. meteorology, building configurations, emissions, etc.)

Assumes homogenous exposure

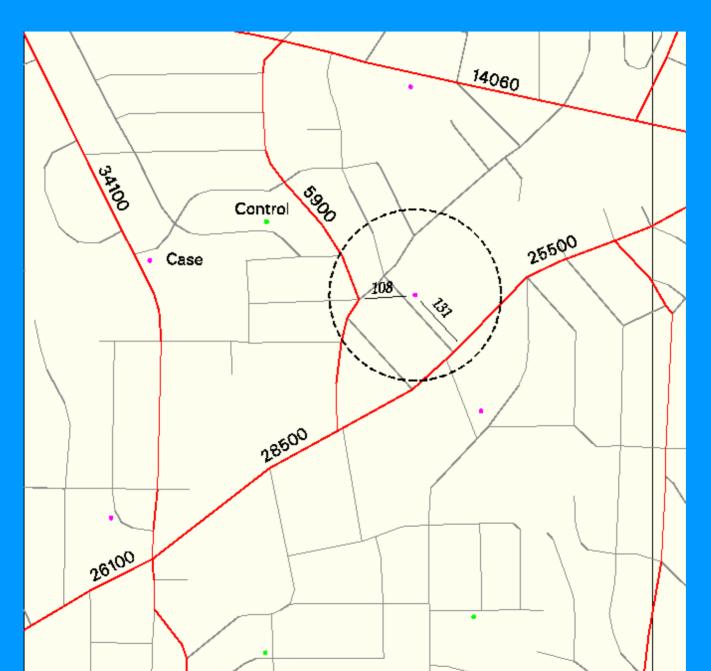
Misclassification (e.g. building heights)

Misclassification (e.g. wind)

Proximity Analysis

- Strengths:
 - Easy to use (adapt with Gaussian weights)
 - Distance of residence to road correlates well with personal and ambient NO₂ monitoring (Rijnders, et al 2001)
 - Weaknesses:
 - Exposure misclassification likely without wind direction data
 - Does not model actual level of pollutants

Proximity Analysis



Dispersion Models (e.g. CALINE)

• Strengths:

- More accurately measuring dispersal of pollutants
- Model actual pollutant levels

- Weaknesses:

• Is gaussian plume realistic model?

Surface showing dispersal of hydrocarbons - light (high exposure) to dark (low exposure) each pixel has grams hydrocarbons per 10 meters



Ν

Land-use regression (e.g. Briggs)

Strengths:

Use land use, met data, DEMs, traffic to predict pollutant concentrations
Easily obtained data

Weaknesses:

 Need enough monitoring locations for calibration/validation

 Need to replicate in new areas; models developed in one geographic location may not be predictive in other areas.

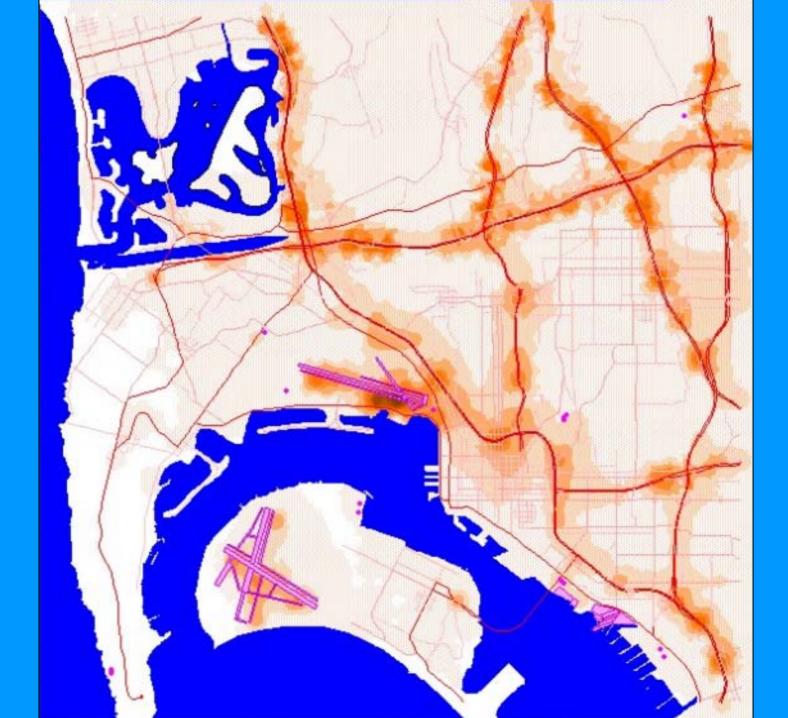
Integrated Meterological-Emission Models (e.g. ADMS-Urban, Cal-PUFF)

– ADMS-Urban: Strengths:

- Incorporates mobile, point, and area sources
- Ability to model gridded emissions and terrain simultaneously
- model 10,050 receptor points
- boundary layer effects and dispersal behaviors over complex terrain
- photochemistry.
- seamlessly integrated with ArcView GIS

Integrated Meterological-Emission Models (e.g. ADMS-Urban, Cal-PUFF)

- ADMS-Urban:
 - Weaknesses::
 - Significant training and expertise necessary
 - Cost
 - Multiple data inputs
 - What is the bang for the buck?



Incorporating Time Activity/Personal Monitoring into Tracking

- Cost/sample size makes these activities prohibitive for tracking
 - Subsample analysis?
 - Survey data on time activities
 - Commuting important exposure time:
 - Use of Transportation demand models?

Specs: 128 bits ROM Frequency: 2.45 GHz



*Size compared to a grain of rice

Conclusions

- Integrating density estimation techniques important for visualization and analysis of health tracking data – increased method development necessary
- Various approaches for traffic-exhaust modeling for tracking – Need to capture most accurate method over the most population at least cost