

LandScan Population Research Program

Presented by

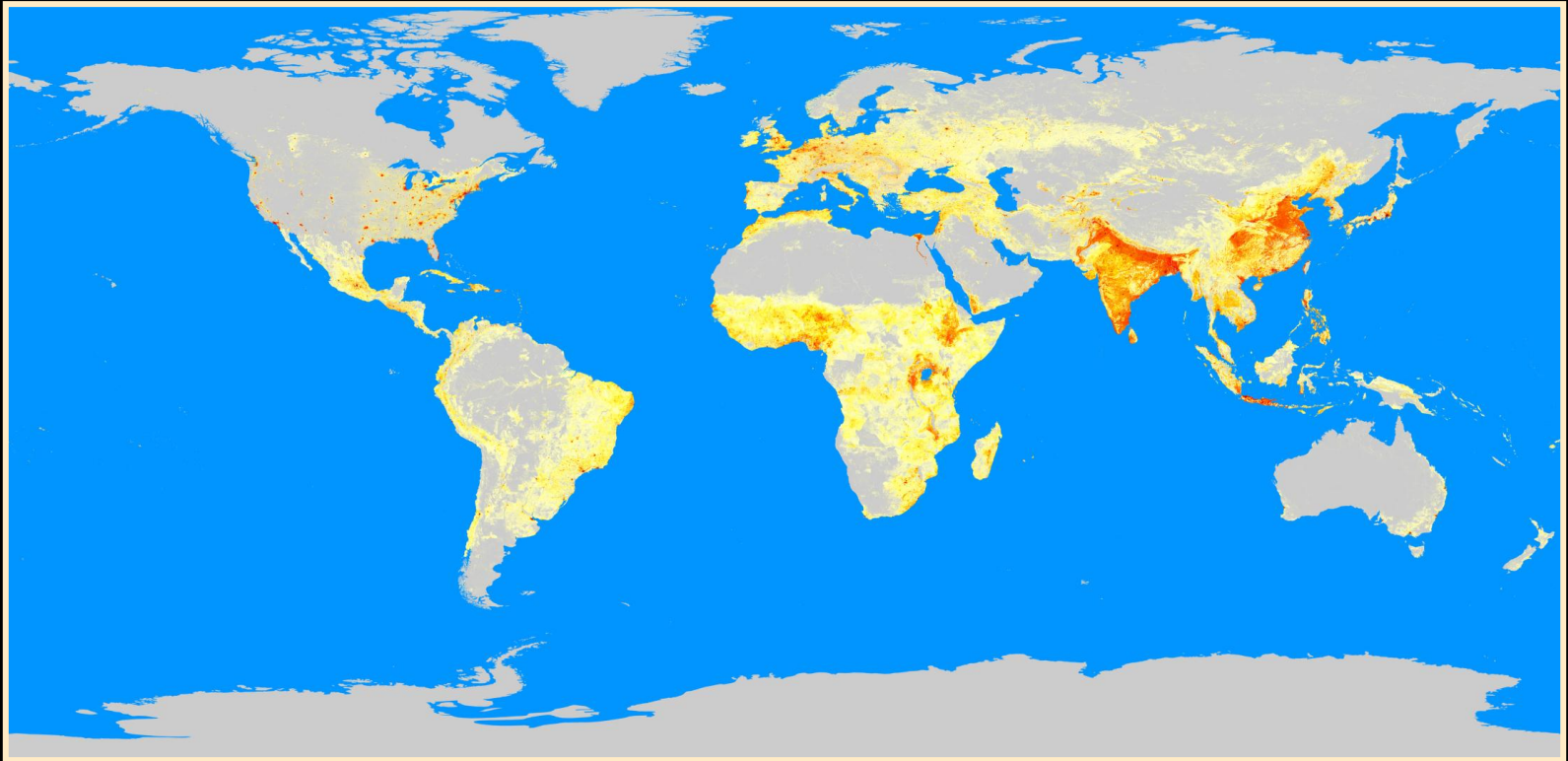
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Geographic Information Science
and Technology
Computational Sciences and Engineering



What is LandScan?

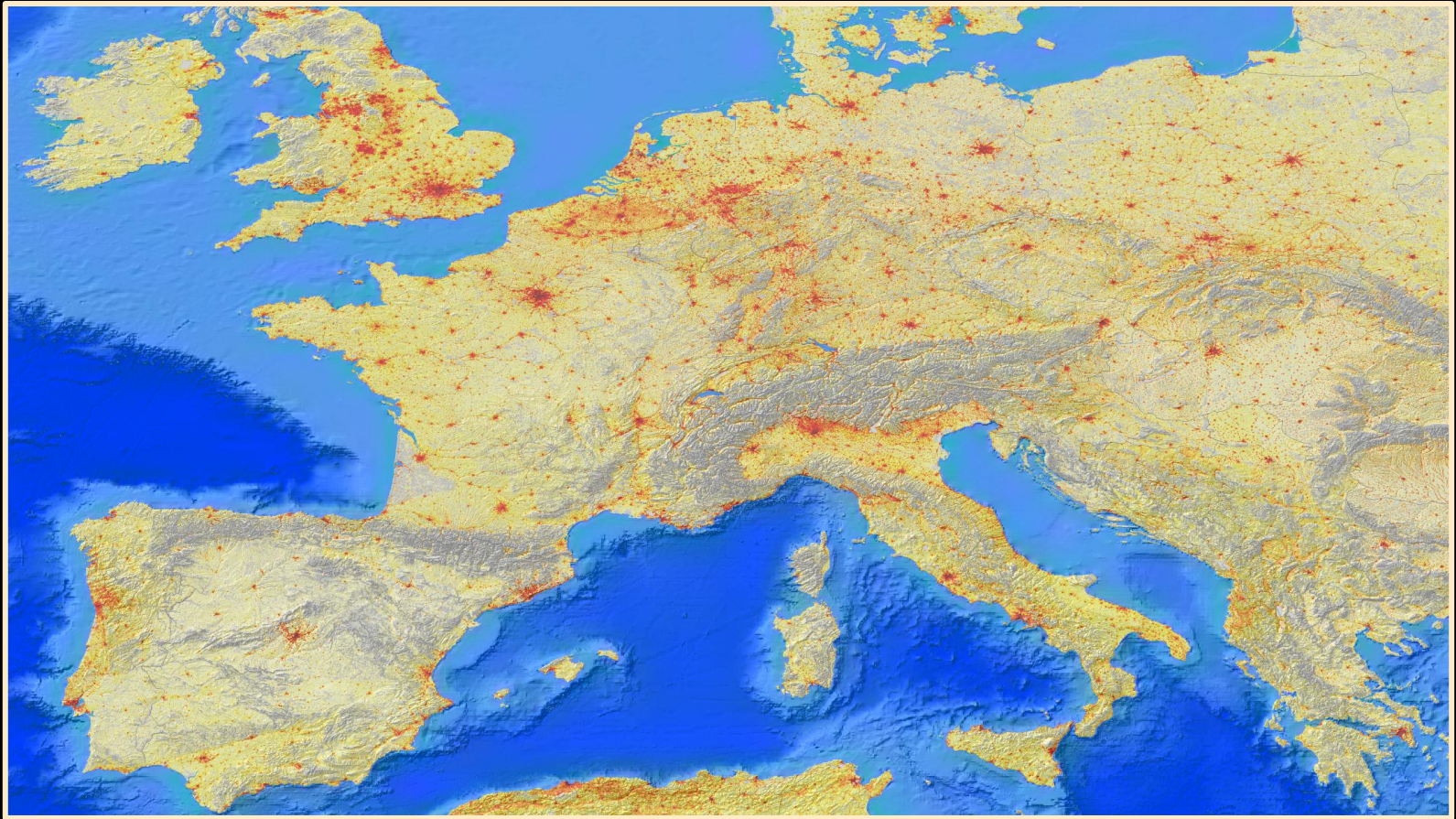
Population distribution model, database, and tool developed from census and other spatial data using a uniform regular grid



Improving knowledge of where people are located

What is LandScan?

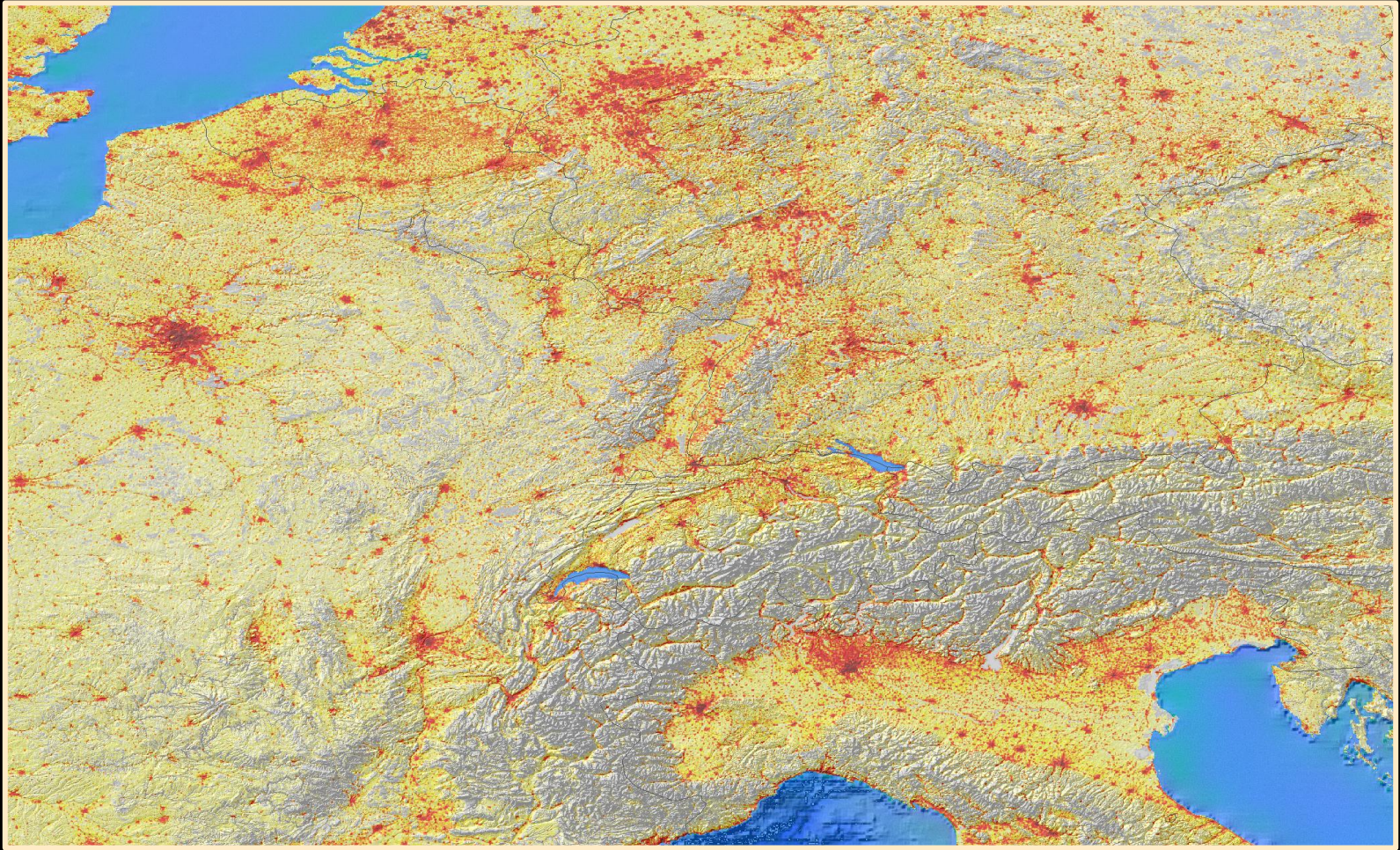
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Improving knowledge of where people are located

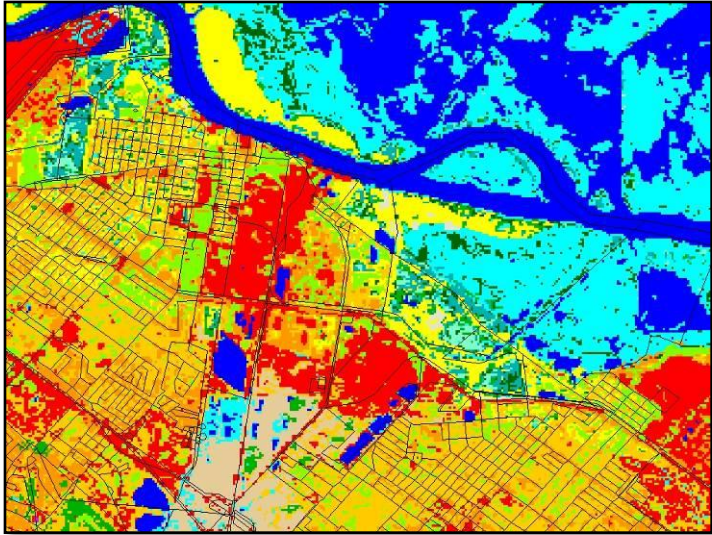
What is LandScan?

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Improving knowledge of where people are located

How is LandScan developed?



- Dasymetric spatial modeling
- Distribute best available census counts to LandScan cells based on a likelihood coefficient calculated by spatial models
- Model structure is the same everywhere, but weights for each variable are tailored to each country
- Similar operations performed for each data layer, and outputs are mathematically combined
- Population is allocated to each cell

H	H	H	L	L	F
H	H	H	L	F	W
H	L	L	L	F	W
H	L	F	W	W	W
L	L	F	W	W	W
L	F	F	W	W	W

H: High-density residential
L: Low-density residential
F: Evergreen forest
W: Water

150	150	150	90	90	4
150	150	150	90	4	0
150	90	90	90	4	0
150	90	4	0	0	0
90	90	4	0	0	0
90	4	4	0	0	0

Aggregated to coefficient cell size

	1230			372	
	526			0	

325	63
112	0

$$Population_{Cell\ i,j} = PC_{Block} \times W_{cell\ i,j}$$

3250	630
1120	0

$$PC_{Block} = \frac{Total\ Population_{Block}}{\sum_1^n W_{Cell\ i,j}}$$

Product of additional data types (e.g. distance from roads, slope)

$$W_{Cell\ i,j} = LC_{i,j} \times PR_{i,j} \times S_{i,j} \times LM_{i,j}$$

Annual improvements

- Census updates
- Higher resolution input data
- High-resolution imagery
- Processing techniques
- Algorithm refinements

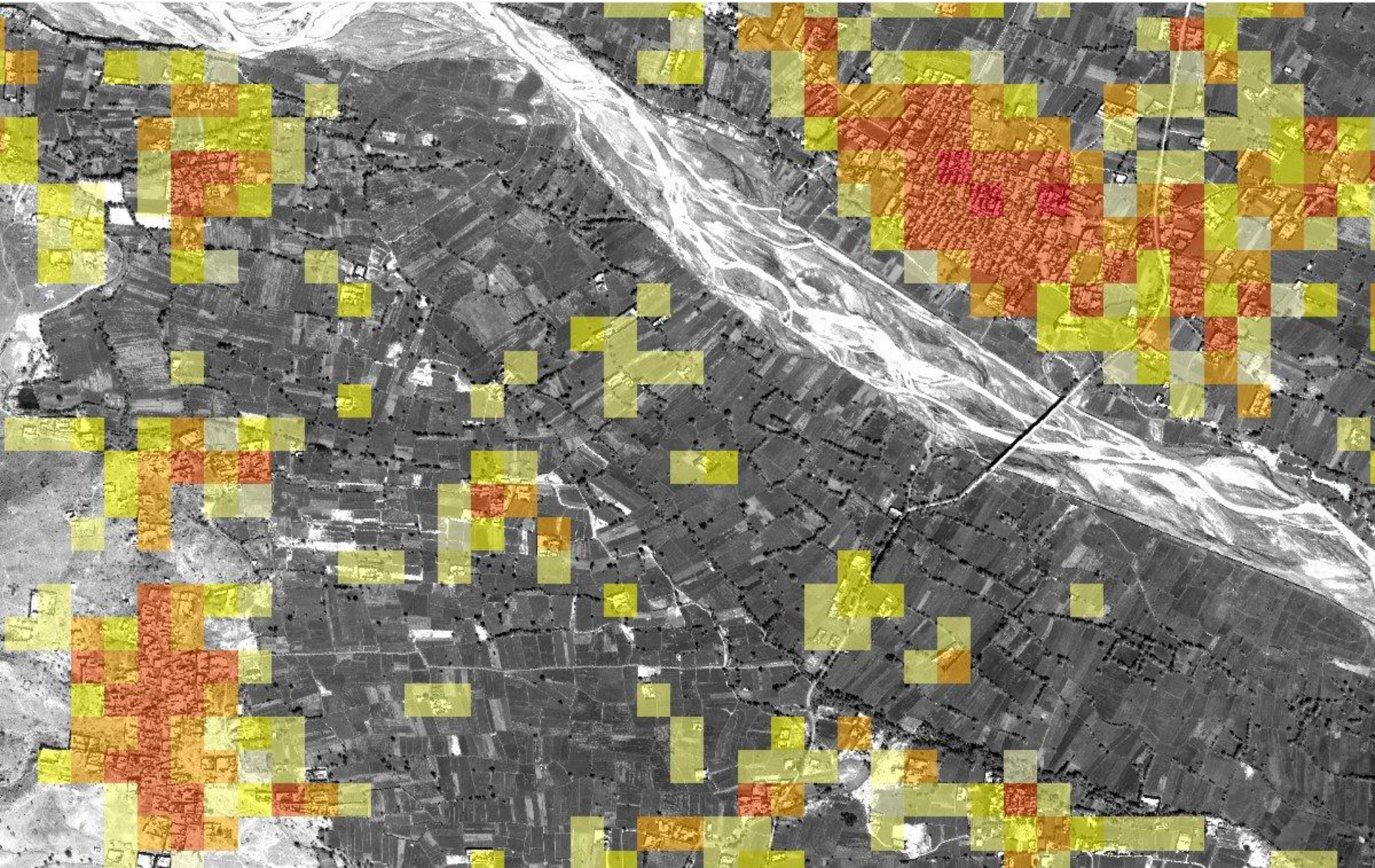
LandScan 2009

Annual improvements

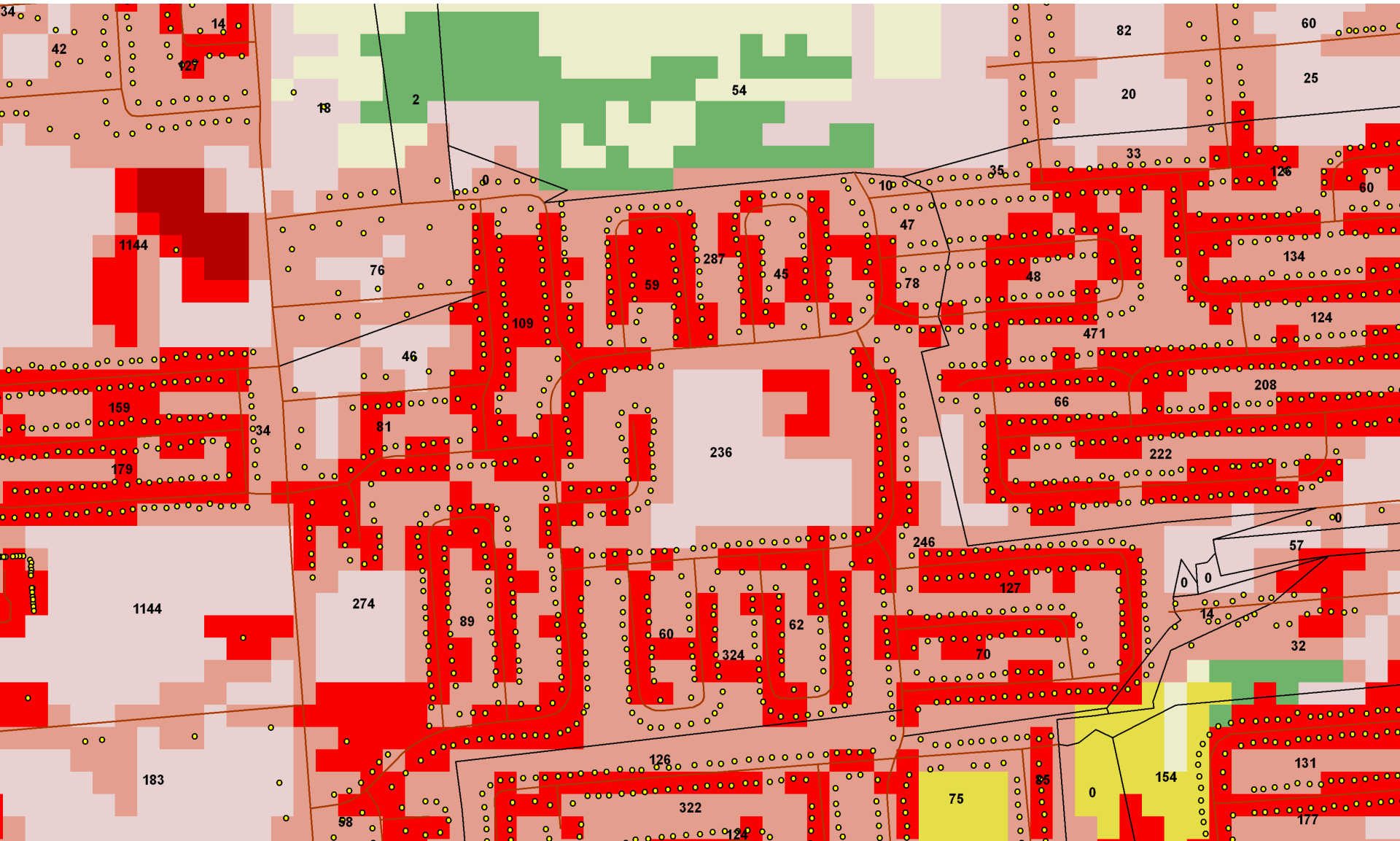
- Census updates
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LandScan 1998

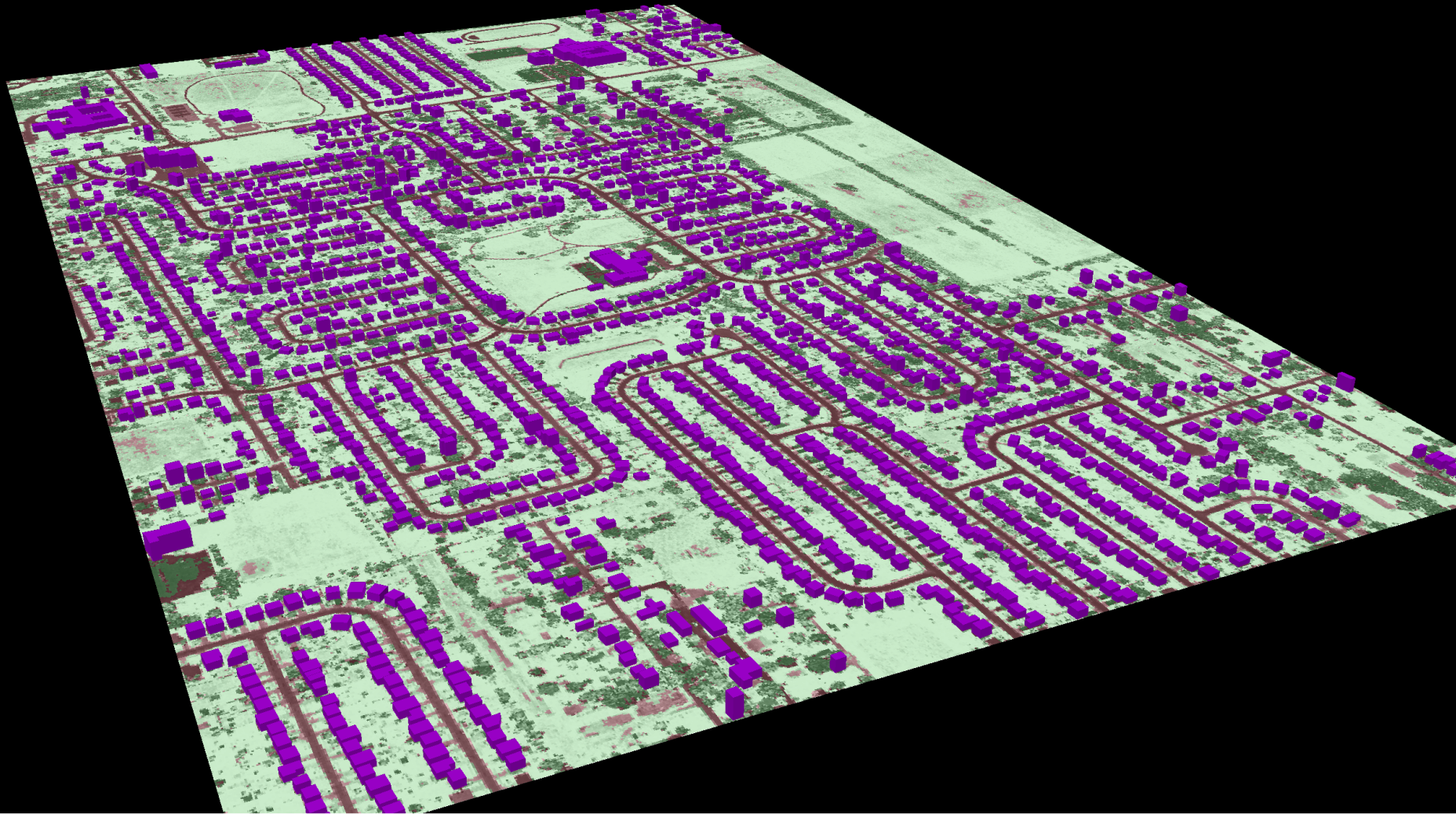
Spatial refinement of LandScan Global



Address point locations



LIDAR data with building elevations



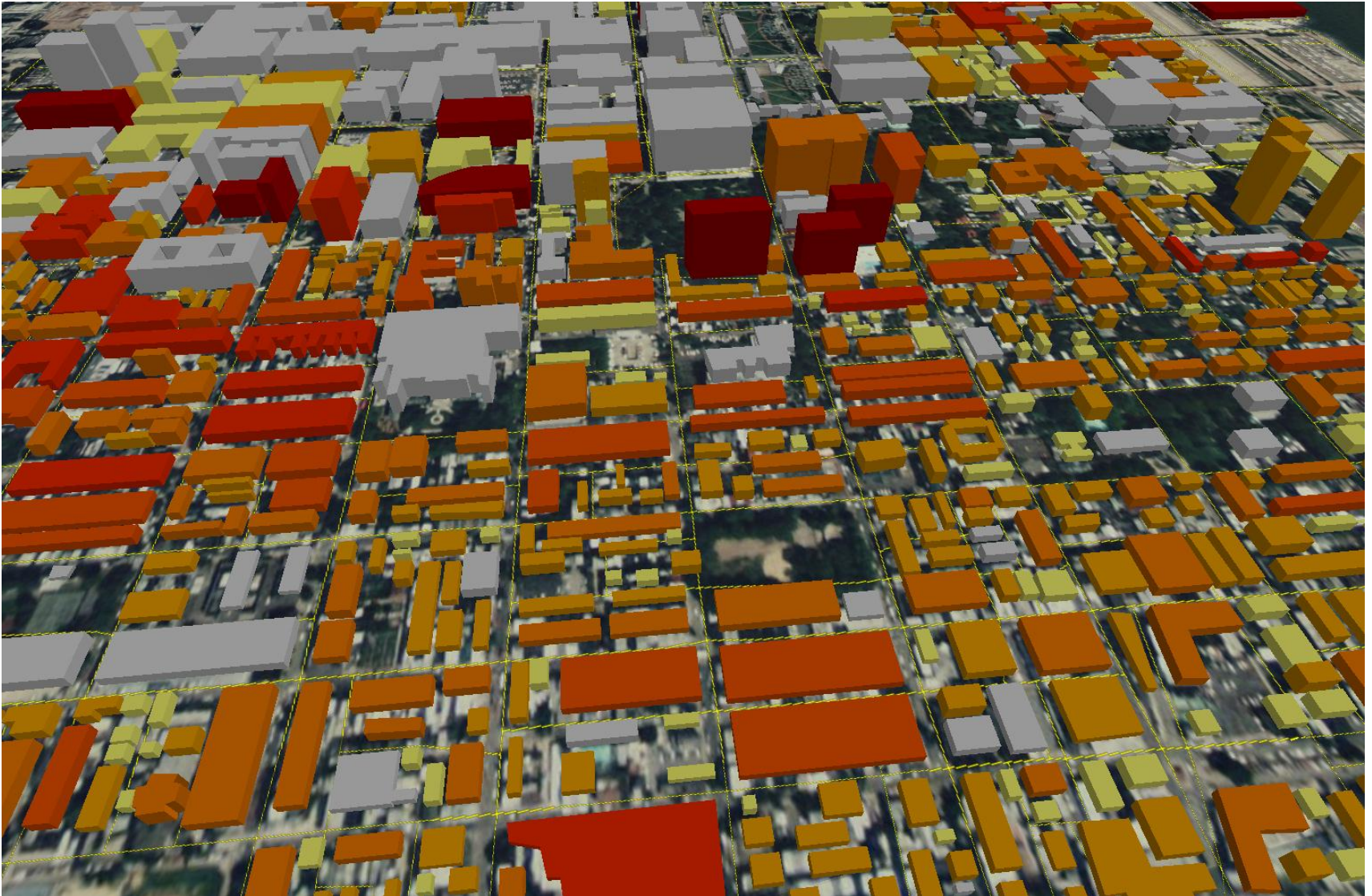
Lidar data with building extractions



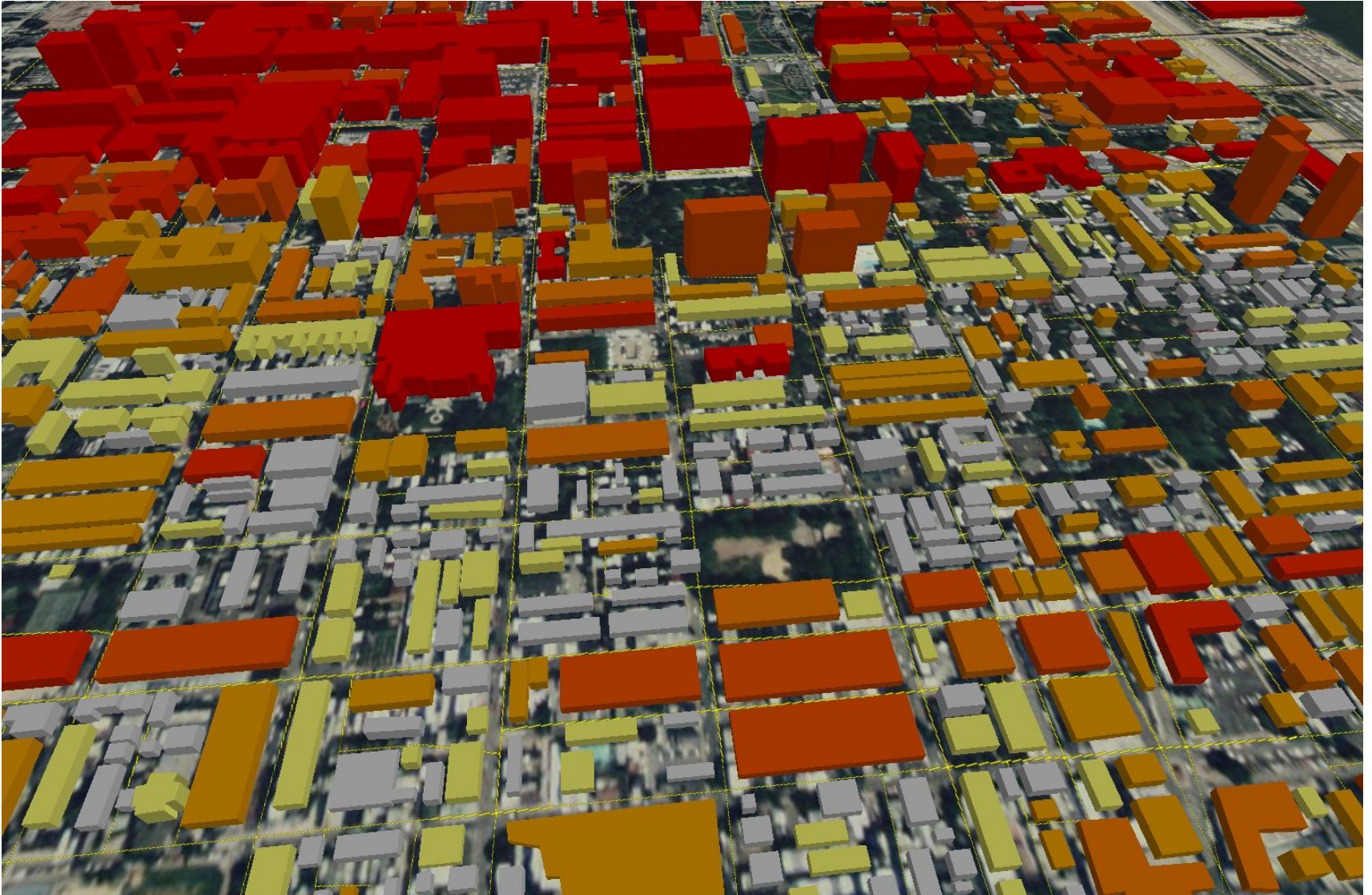
Lidar data with building extractions



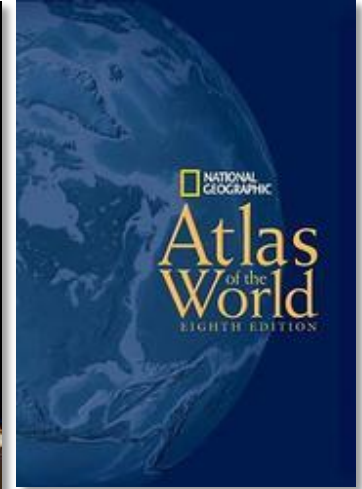
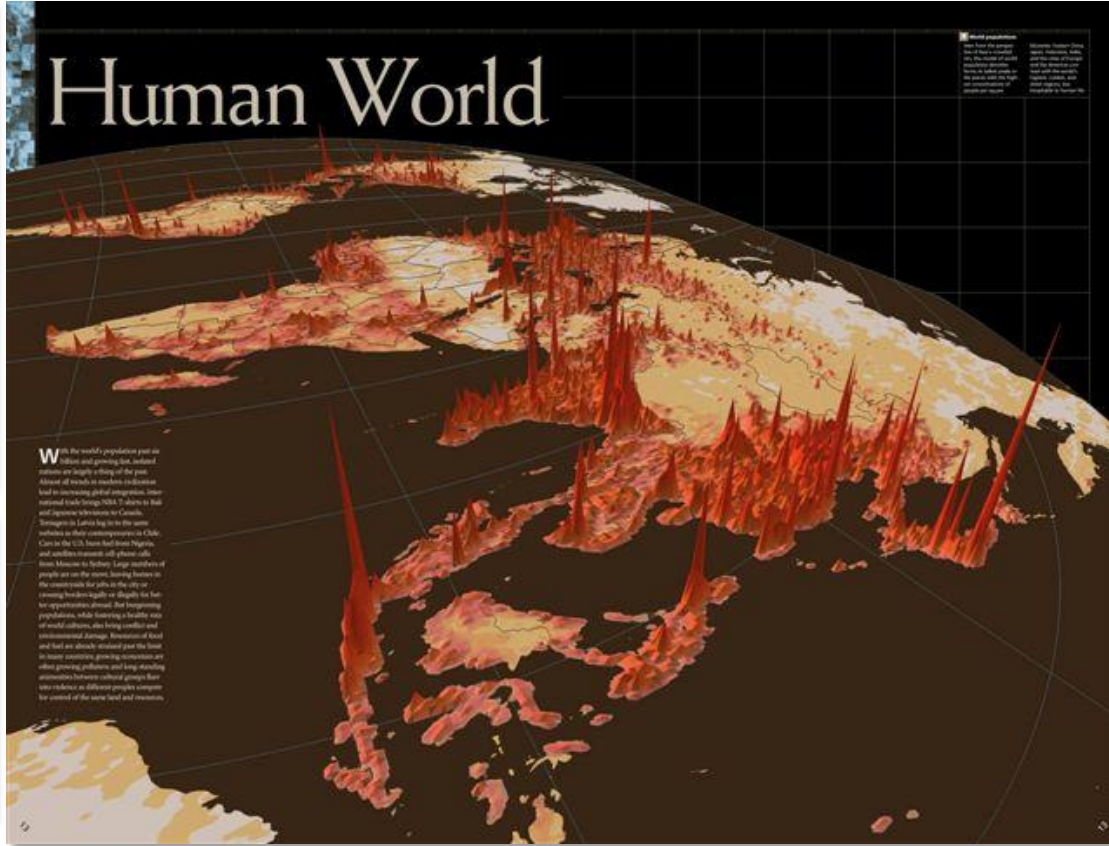
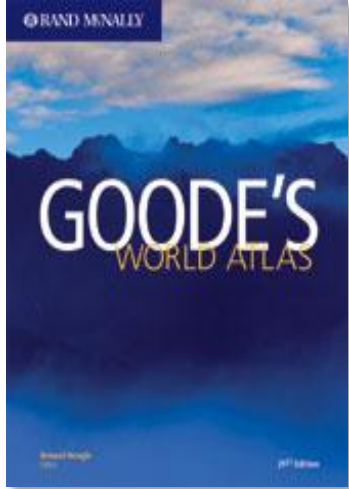
Residential population



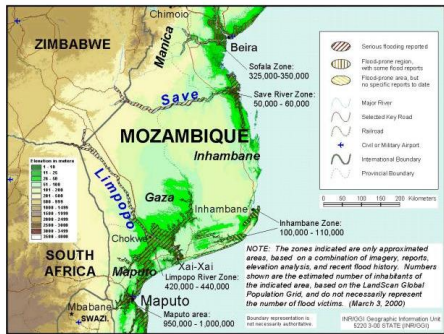
Workers population



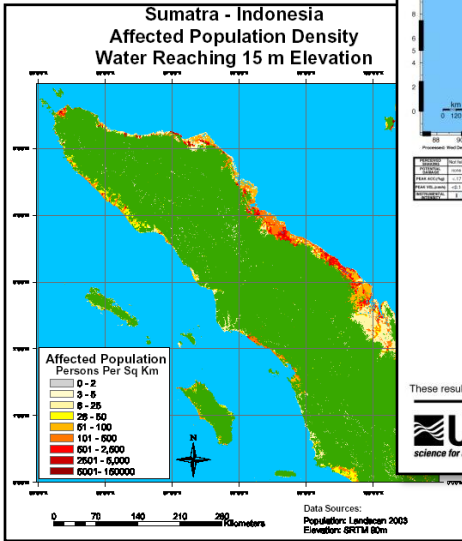
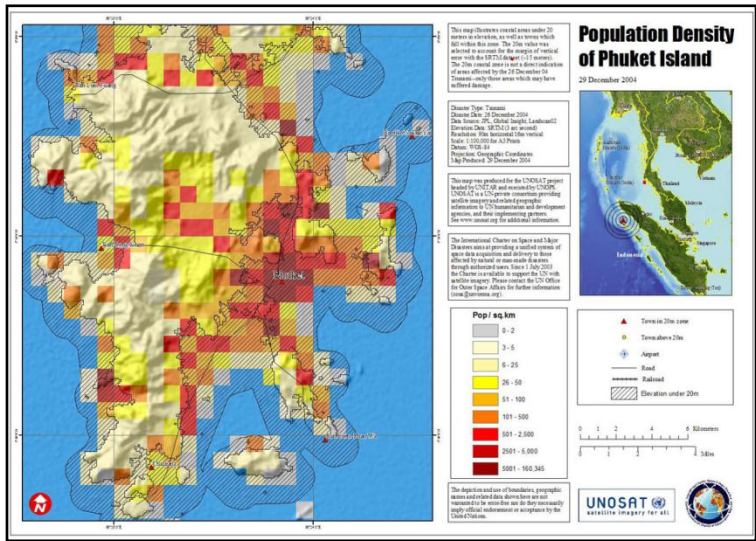
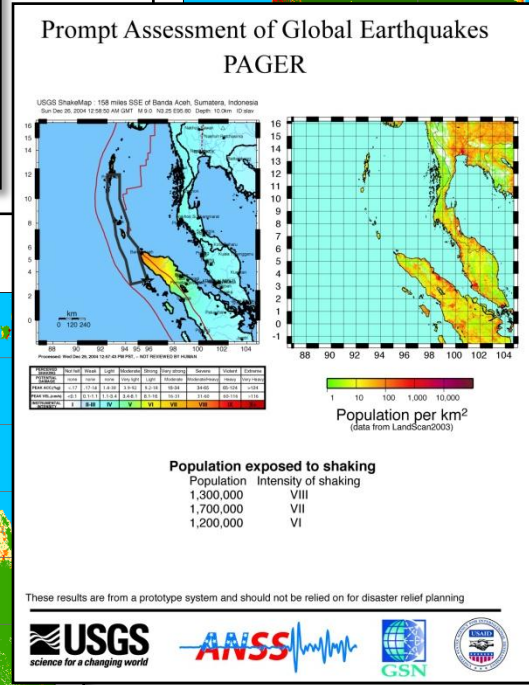
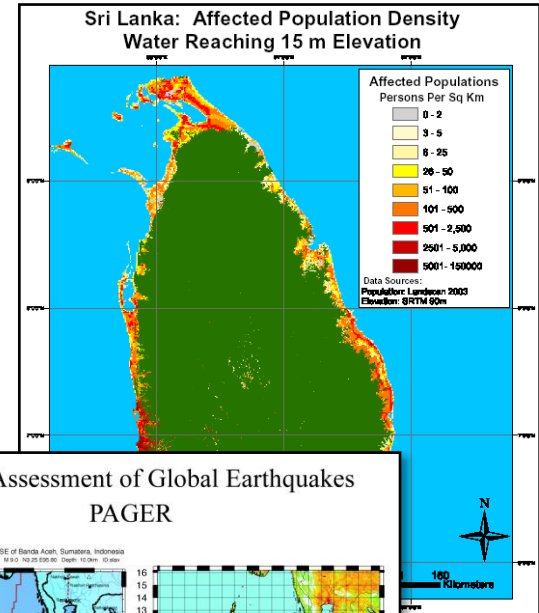
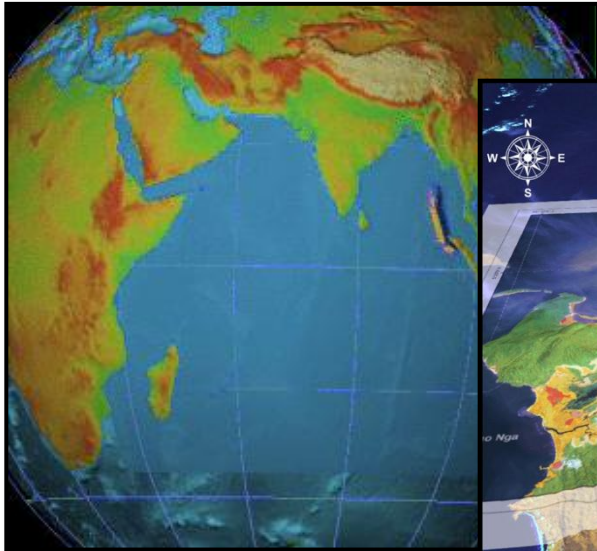
Community standard for population



MOZAMBIQUE FLOOD ZONES (Probable Areas)



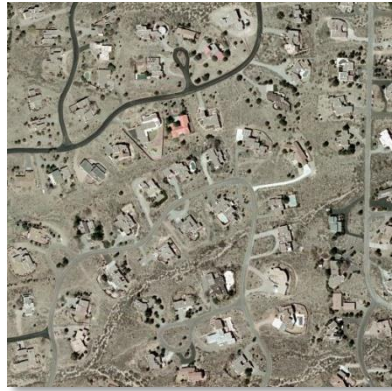
LandScan application: Tsunami relief







Patterns in overhead imagery



Higher Income

Middle Income

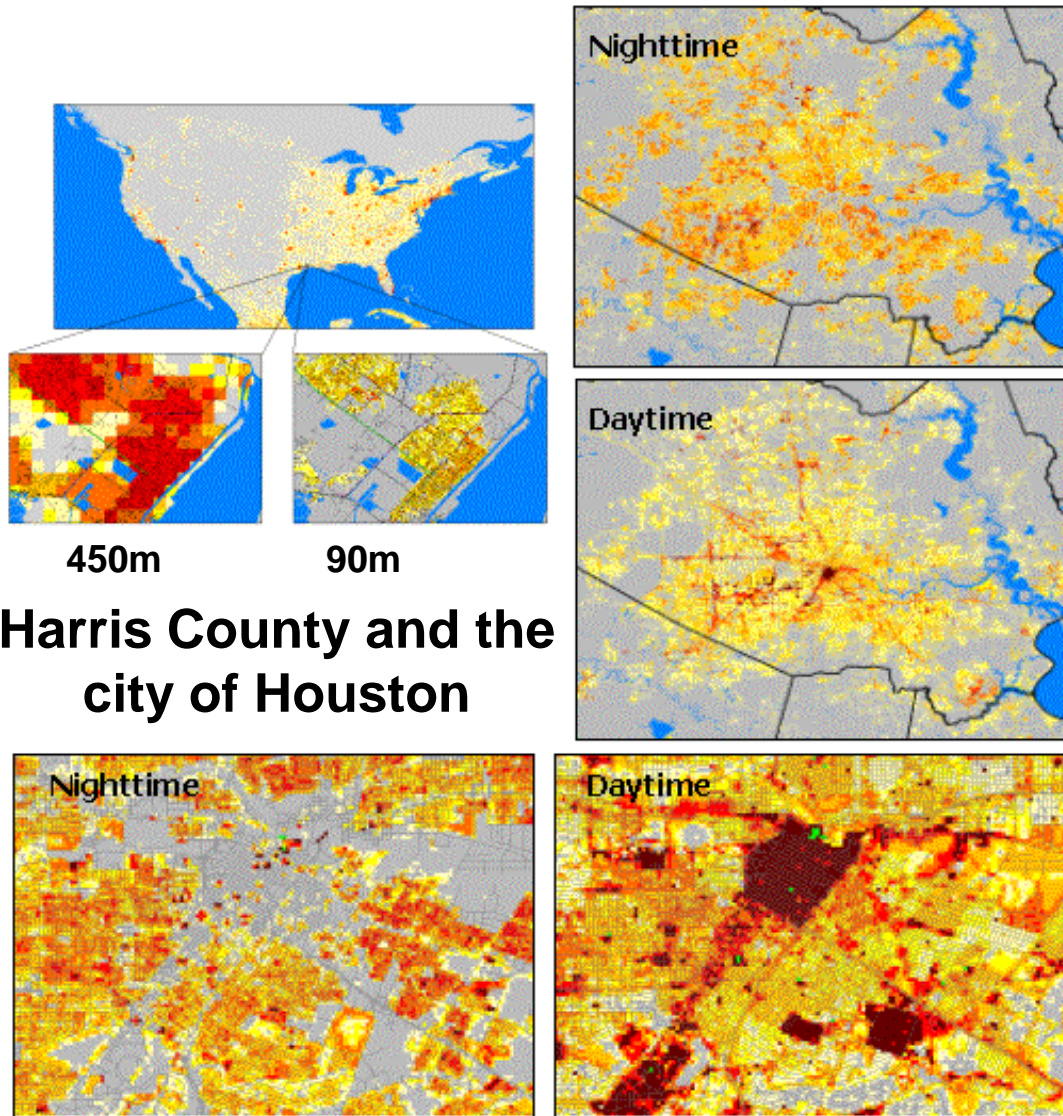


Lower Income

Urban Characterization from UAV Stream

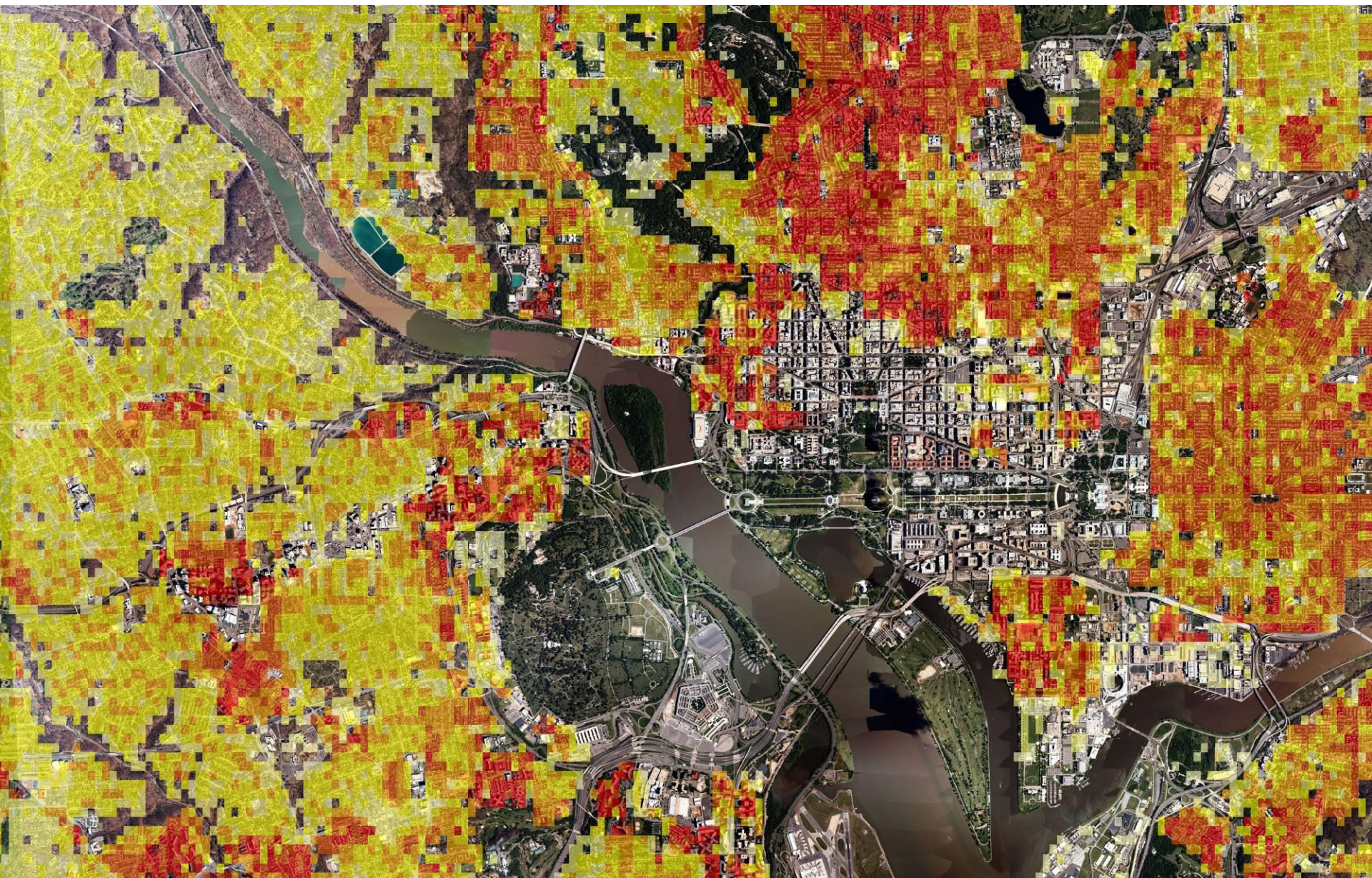


LandScan USA

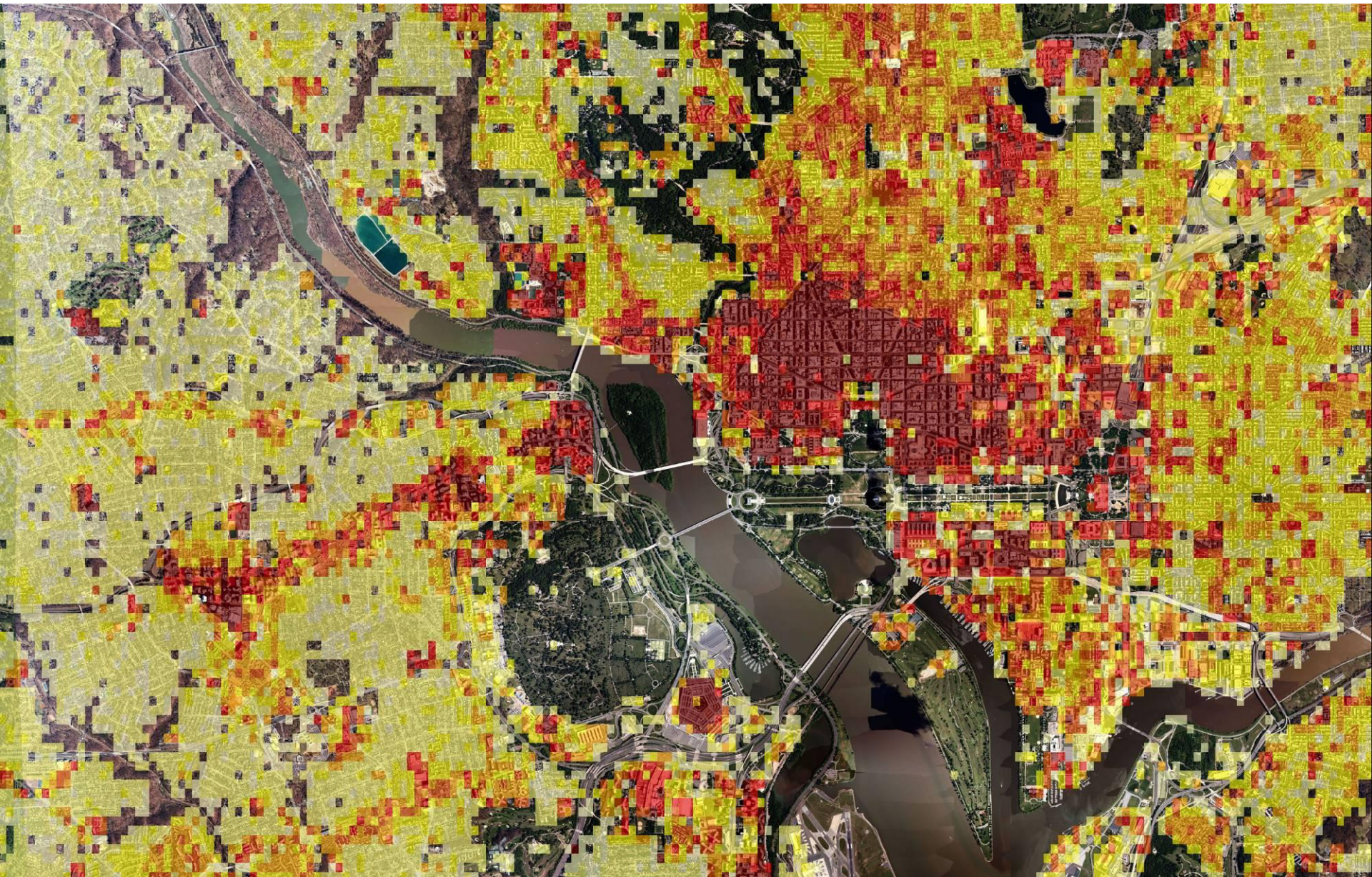


Harris County and the city of Houston

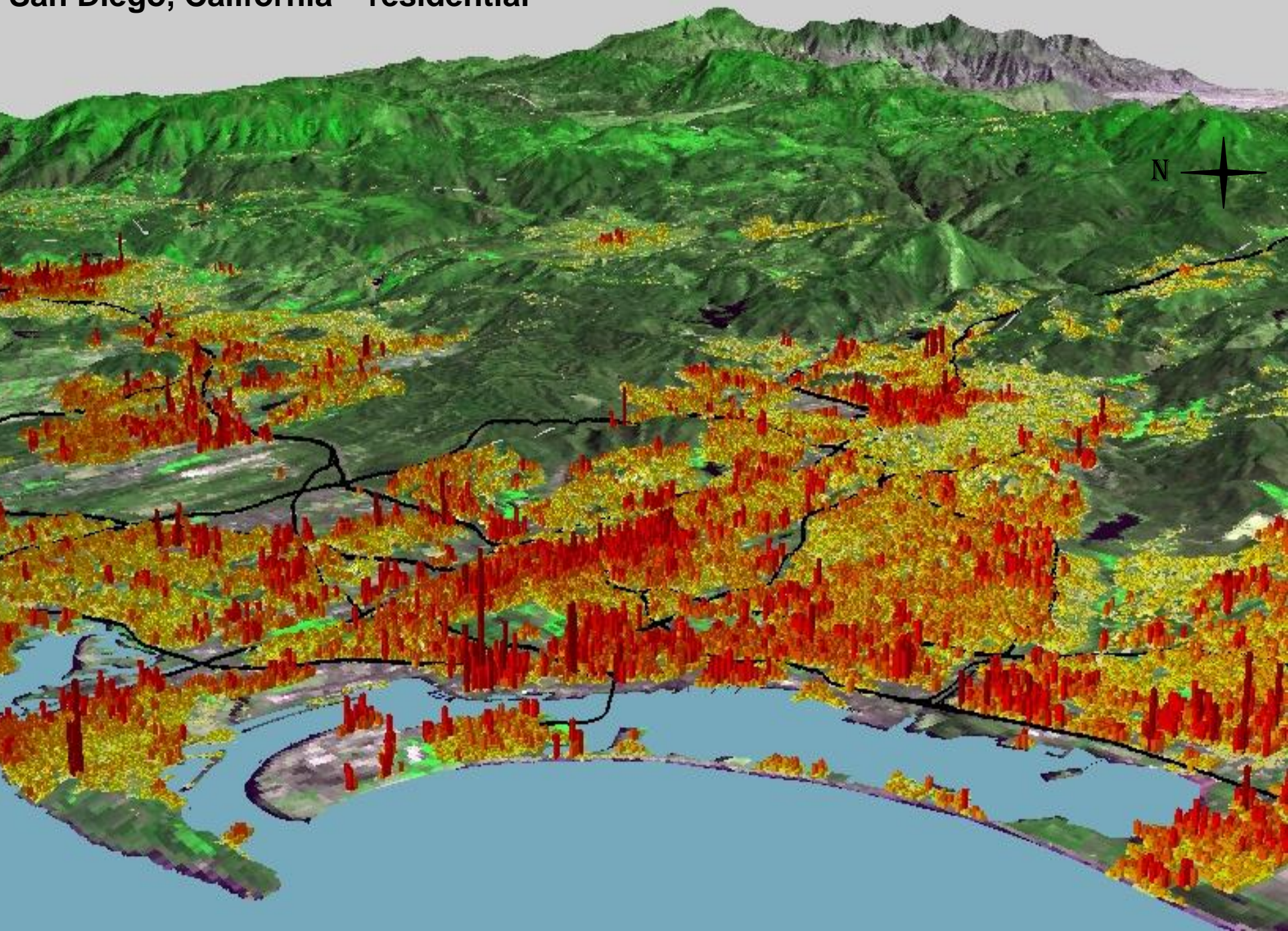
Very high spatial resolution	3 arc second resolution (about 90 m or 300 ft) or finer
Finer temporal resolution	Nighttime (residential) and daytime population distributions
High currency	Modeled from best available census block data
Value added	Demographic and socioeconomic characteristics for easy integration with risk and impact assessment models
Compared with census data	Better than census resolution for most city blocks



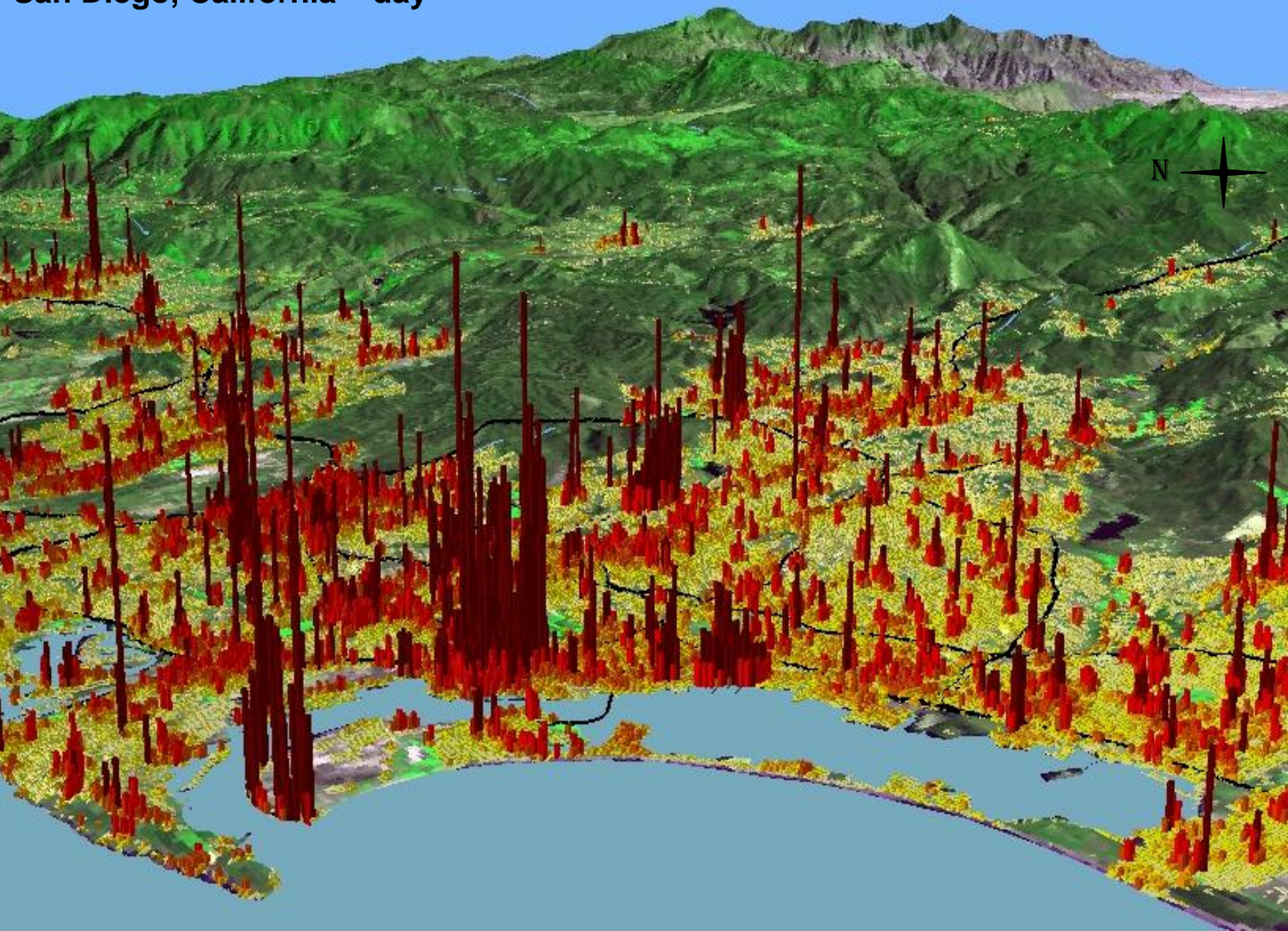
Daytime distribution



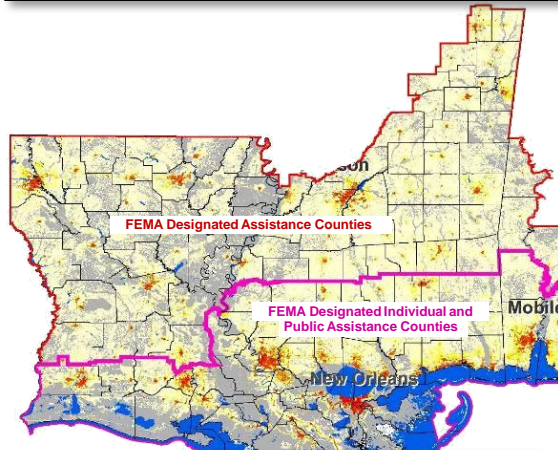
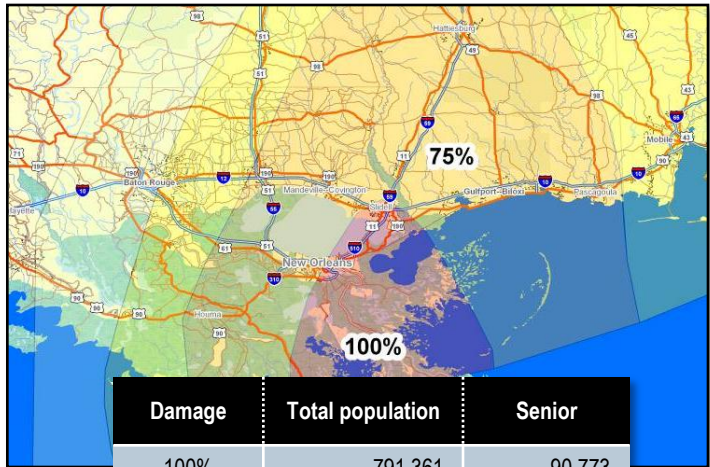
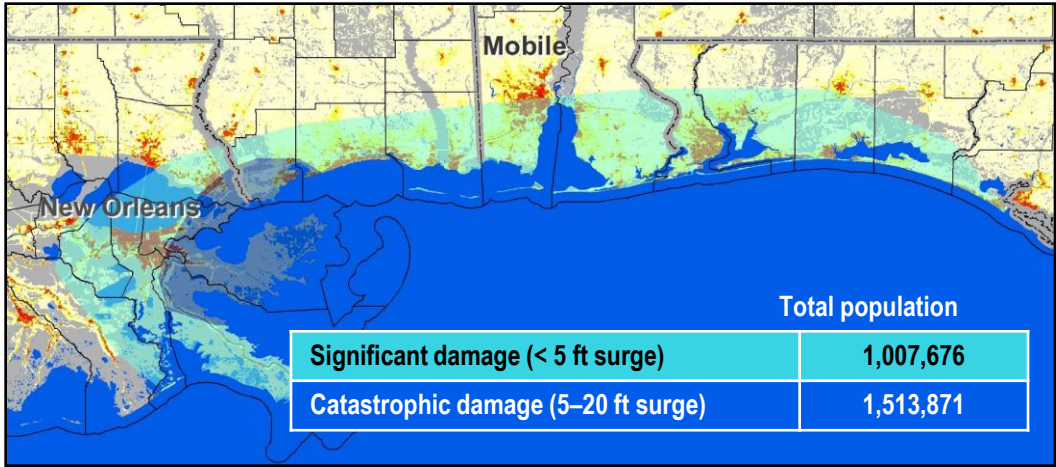
San Diego, California – residential



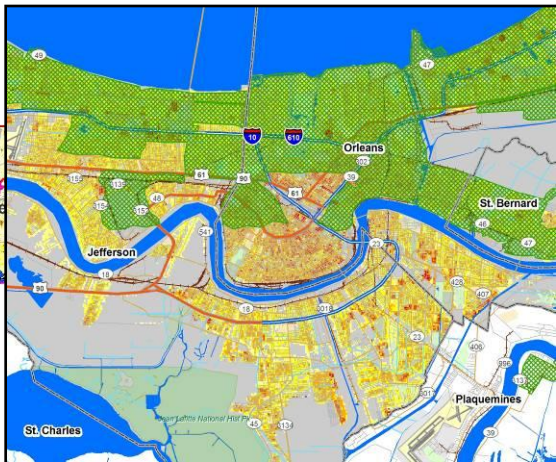
San Diego, California – day



Hurricane preparedness and response

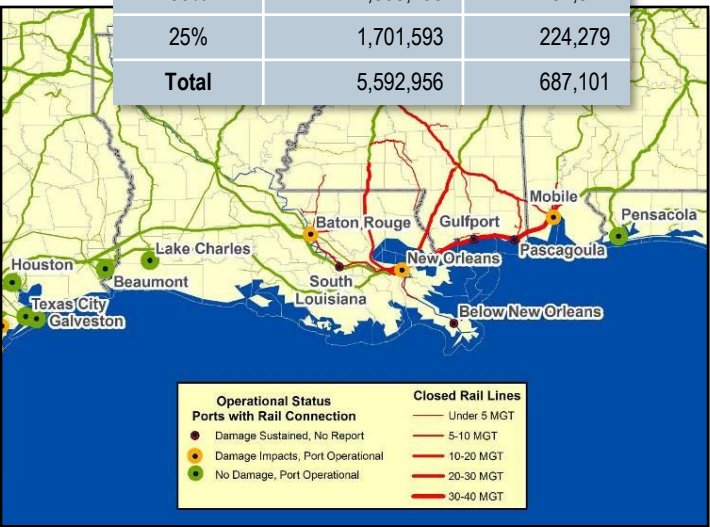


	Individual and public assistance	Public assistance
Alabama	575,133	56,801
Mississippi	707,506	1,391,233
Louisiana	3,153,293	1,362,477



FEMA Impacted Areas

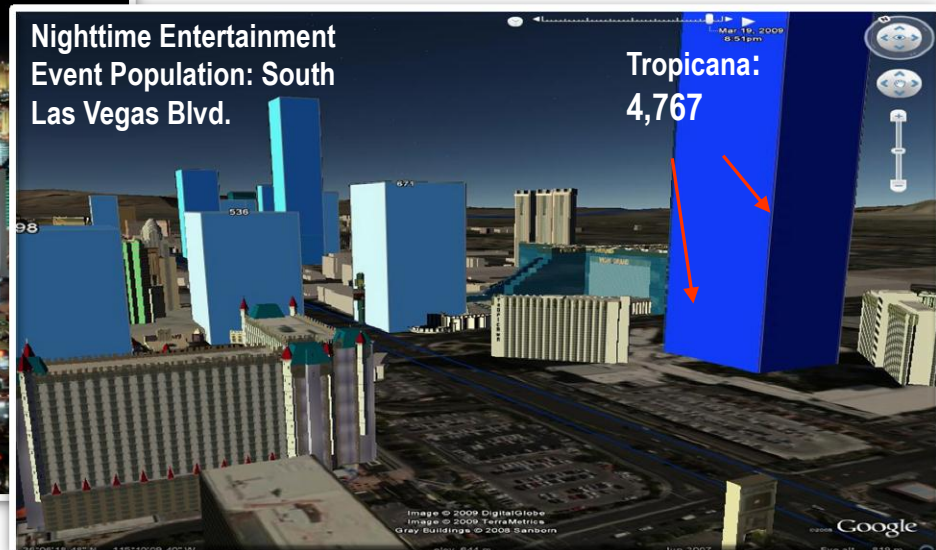
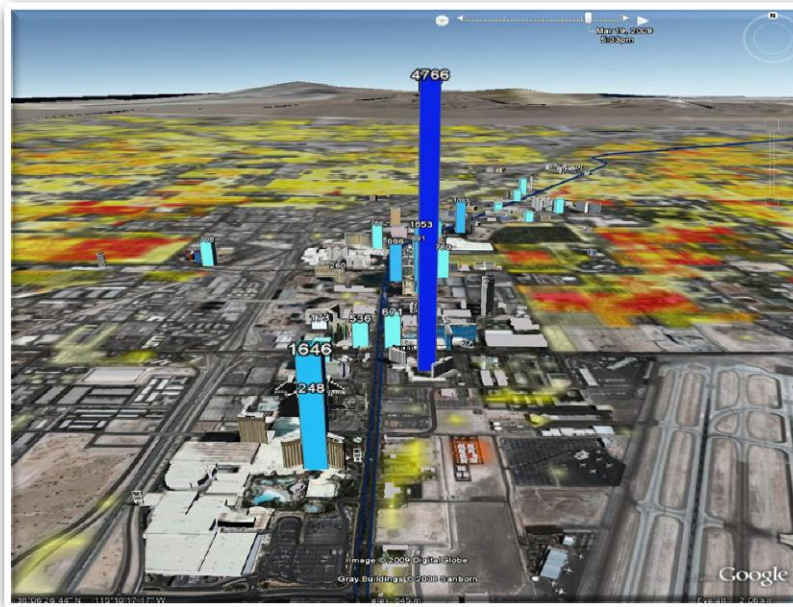
Source: FEMA Impacted Areas, August 31, 2005



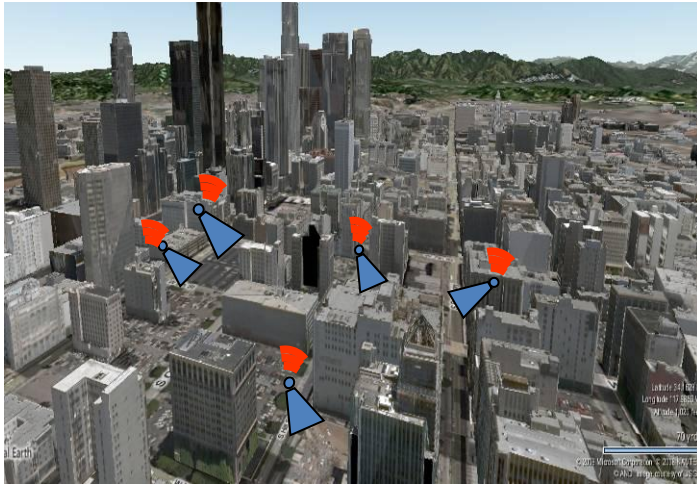


Transient populations

Distribution of nighttime entertainment event population



Automated crowd scene analysis



Problem statement

- Dense crowd situations pose many challenges to automated tracking systems
 - Learning background models in crowded situations and ad-hoc camera views is hard
 - Inter- and intra-object occlusions are highly common. Objects usually occupy only a few pixels
 - Fitting shape and appearance models may be difficult

Technical approach

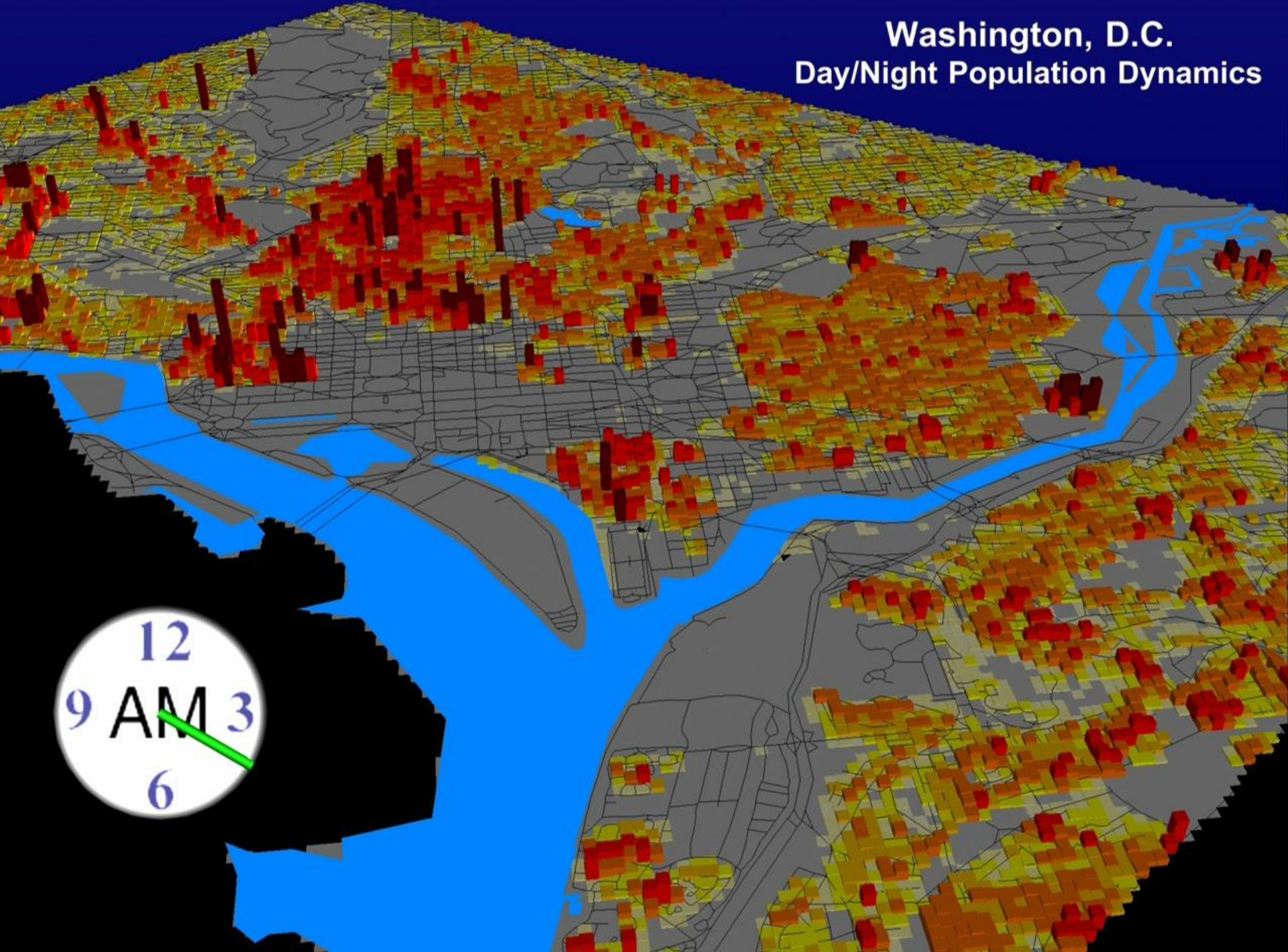
- Identify dominant crowd motions in video where individual tracking is highly challenging. Our approach is based on clustering low-level feature tracks, which may be fragmented or noisy. The similarity between two point tracks is measured using a Longest Common Subsequence based algorithm. The matching cost is computed using dynamic programming

Benefit

- Computer vision system(s) can autonomously interpret crowded environments in a wide range of operating conditions for public safety systems and other persistence surveillance systems

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Washington, D.C. Day/Night Population Dynamics



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