

Context sensitive scaling of energy use simulations

From individual buildings to cities

Andreas Christen ⁽¹⁾

Michael van der Laan ^(1,2)

Ron Kellett ⁽²⁾

Rory Tooke ⁽³⁾


⁽¹⁾ Department of Geography
University of British Columbia

⁽²⁾ School of Architecture and
Landscape Architecture
University of British Columbia

⁽³⁾ Integrated Remote
Sensing Studio
Department of Forestry
University of
British Columbia




a place of mind

An aerial photograph of a city street, likely in Vancouver, Canada, showing a mix of modern glass skyscrapers and older brick buildings. A prominent red overlay is on the right side of the image, containing white text. The street below is filled with cars and a bus, and a Canadian flag is visible on a building in the distance.

Energy use and emissions estimates for entire cities rely mostly on **inventory data**, which is typically only available at aggregate space and time scales.

At the building-scale, sophisticated building energy models (BEM) exist, but require detailed knowledge about the building and its environment.

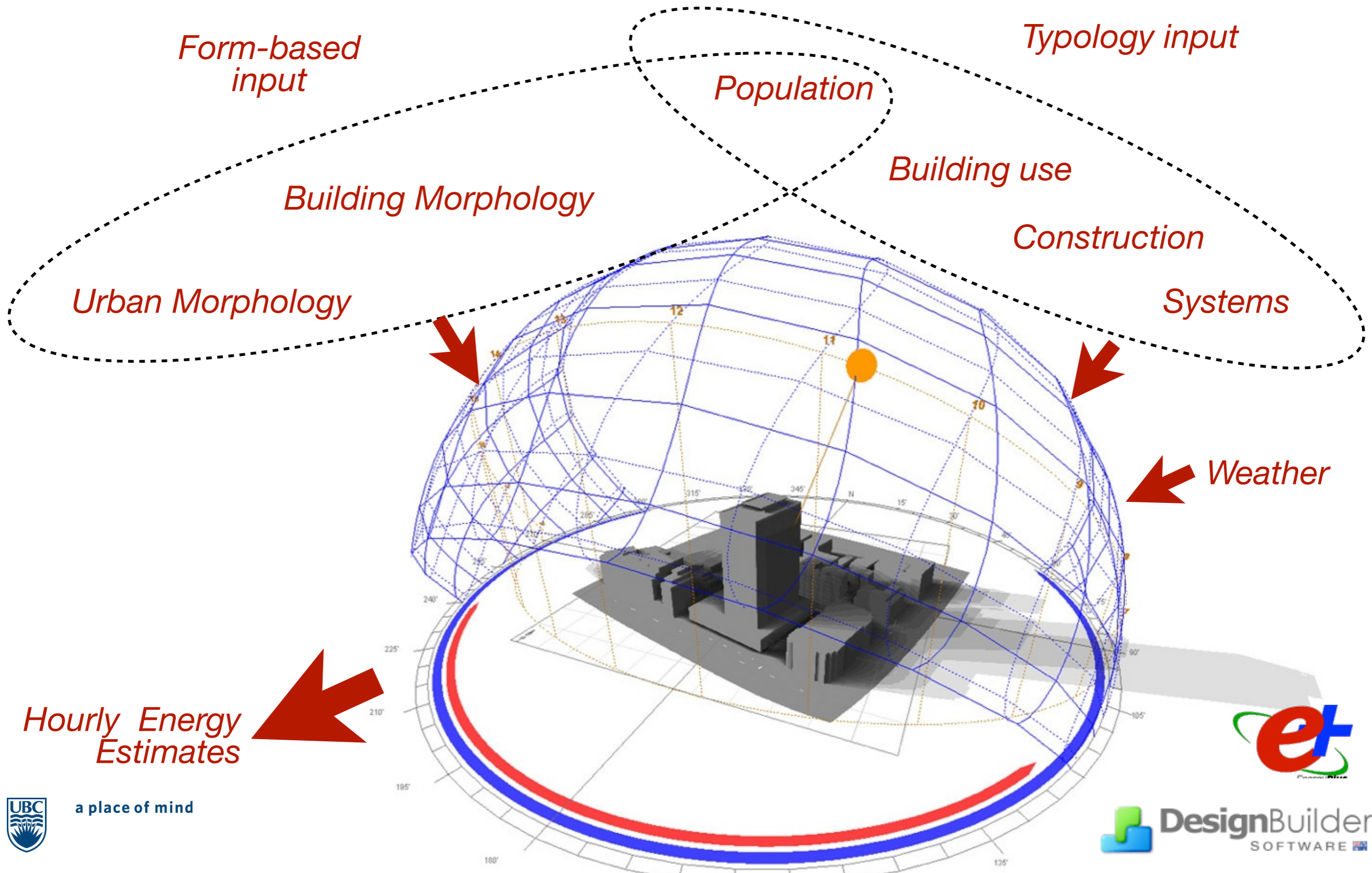


Very few inventories of energy demand are at the **intermediate neighborhood-scale** though this is where many development decisions are made, and information is required for atmospheric models.

This is in large part due to the difficulties in scaling and the effort needed to characterize **urban form**.

Factors influencing energy demand

Potential for automated inputs



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7 km long and 1 km wide LIDAR scan in
Vancouver, Canada
with about 11'000 individual buildings



Central business district
mostly high rises

Suburban
mostly detached
dwellings

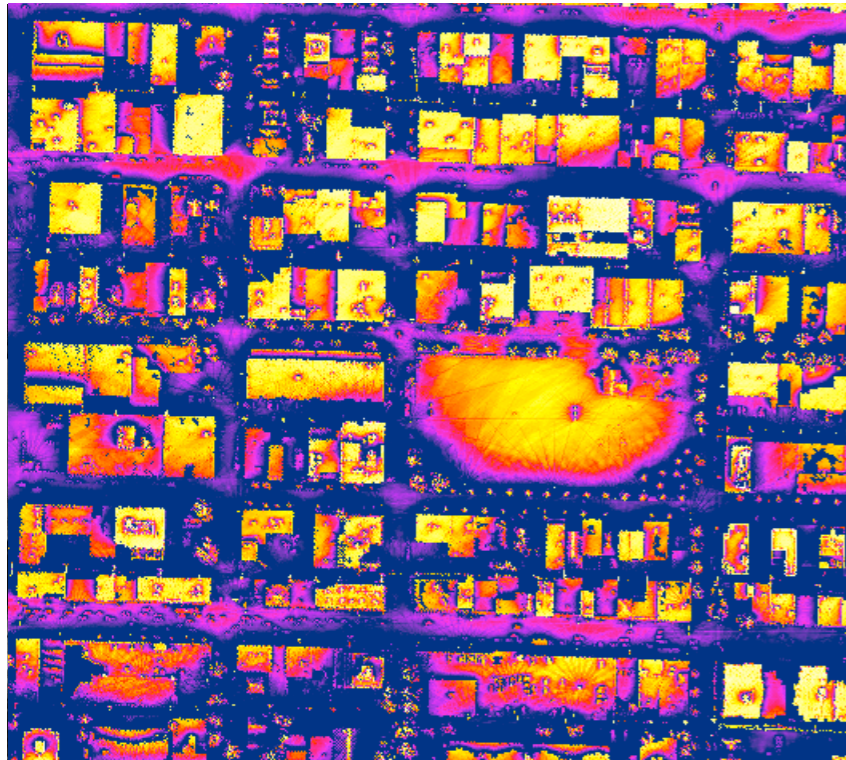


Gridded lidar data products

3D morphometry of buildings and trees at 1 x 1 m

Intersection of Knight and 49th in Vancouver (Goodwin and Coops, UBC, 2008)

What can we derive/inform from LiDAR that is important to building energy use?



Urban morphology
Shading and sheltering



Building morphology
Archetype and form

	LiDAR Volume	avg M3 / capita
Residential Detached (1900-1964)	626	166
Residential Detached (1965-1990)	736	166
Residential Detached (1991-2009)	663	166
Residential Duplex	836	195
Residential Multiplex	983	200
Residential Rowhouse	1205	209
Residential Lowrise / Midrise	2812	206
Residential Midrise / Highrise	36693	223
Mixed-use	1631	772

Population distribution
Hot water demand

Modelling 10'000s of buildings

An archetype approach



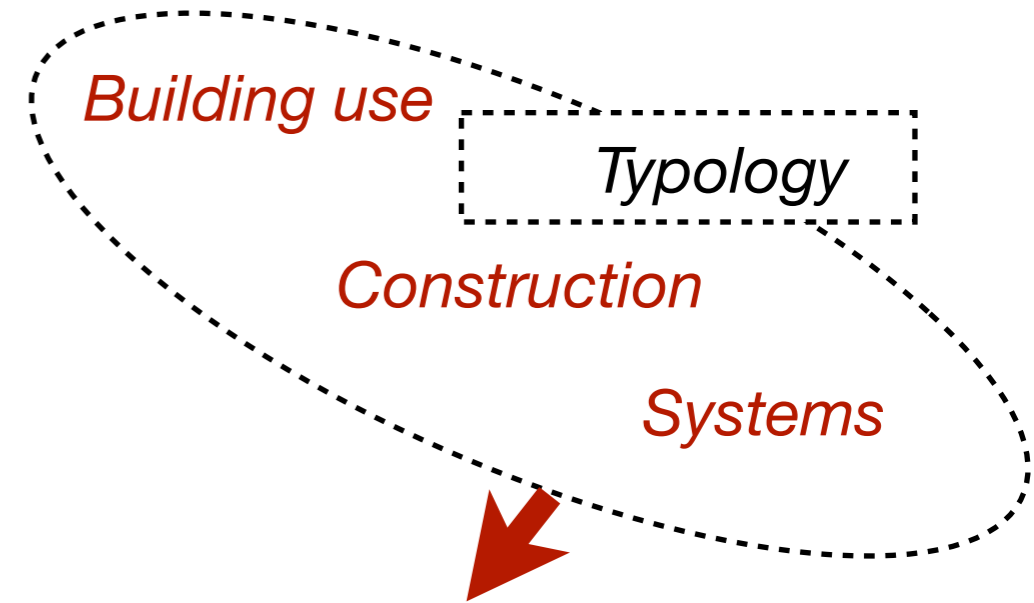
Dalhousie Mixed-use Mid-rise

Apartment housing with exterior circulation above two stories of retail

Case # 33006-00



elements db
BETA



OCCUPIED STORIES

6

DWELLINGS

80

COMMERCIAL UNITS

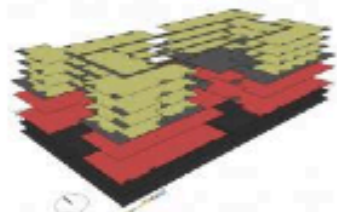
26

APERTURE % of EXTERIOR WALL

0 %

RATIO OF ENVELOPE TO
CONDITIONED FLOOR

0.55



RESIDENTIAL FLOOR AREA

7730 m²

COMMERCIAL FLOOR AREA

7412 m²

TOTAL CONDITIONED

15142 m²

TOTAL PARKING

9875 m²

INDUSTRIAL FLOOR AREA

0 m²

CIVIC FLOOR AREA

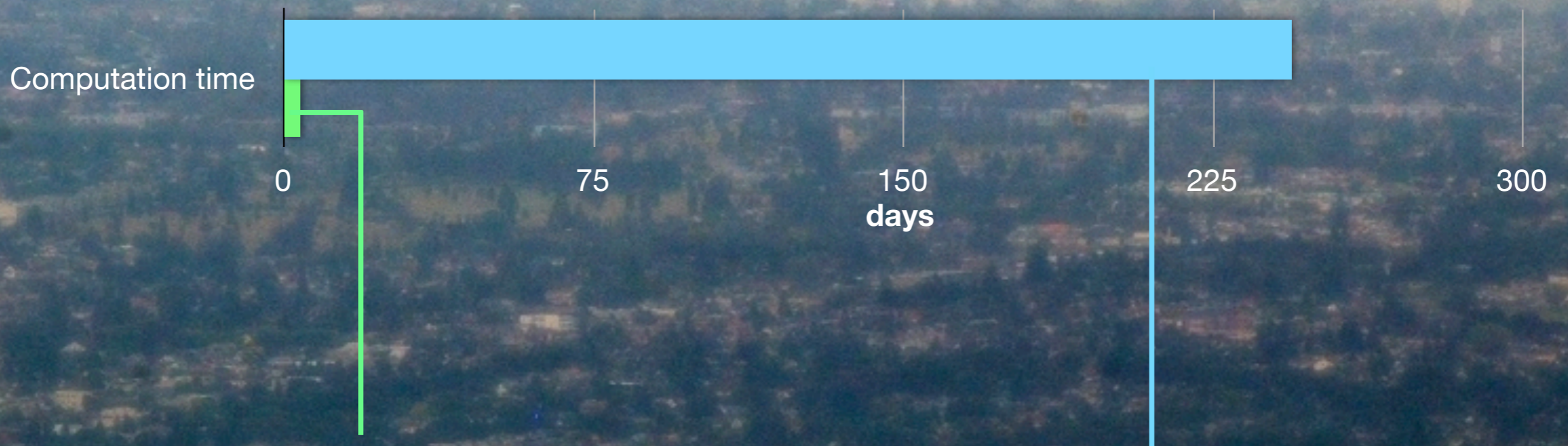
0 m²

Assumes archetypes are
scalable and replicable

12 Archetypes chosen based on an assumed correlation between building use, form, period of construction and energy performance



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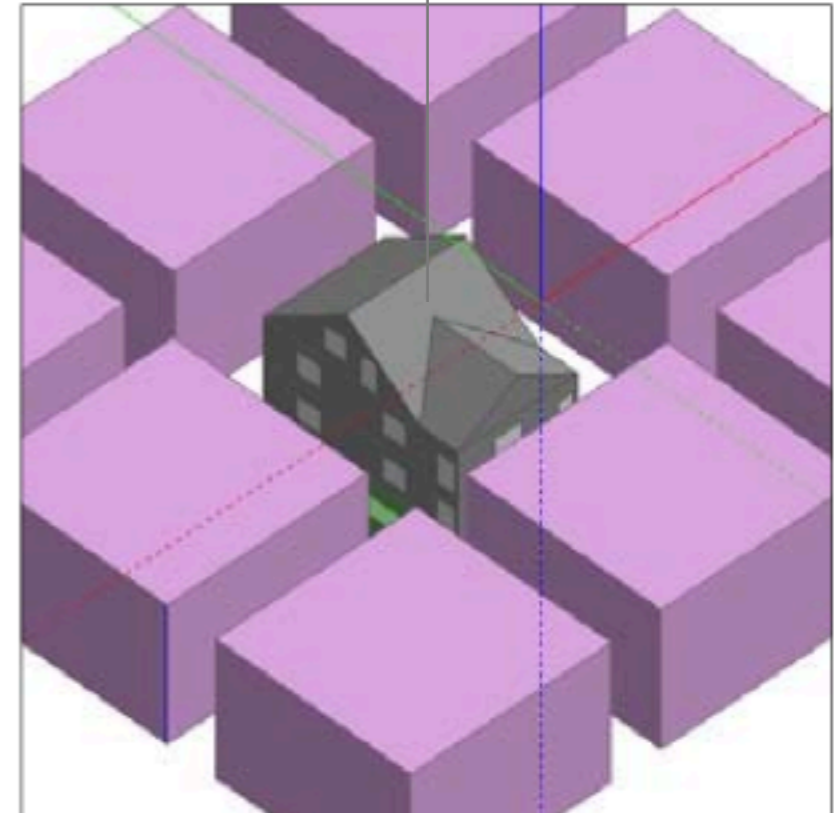
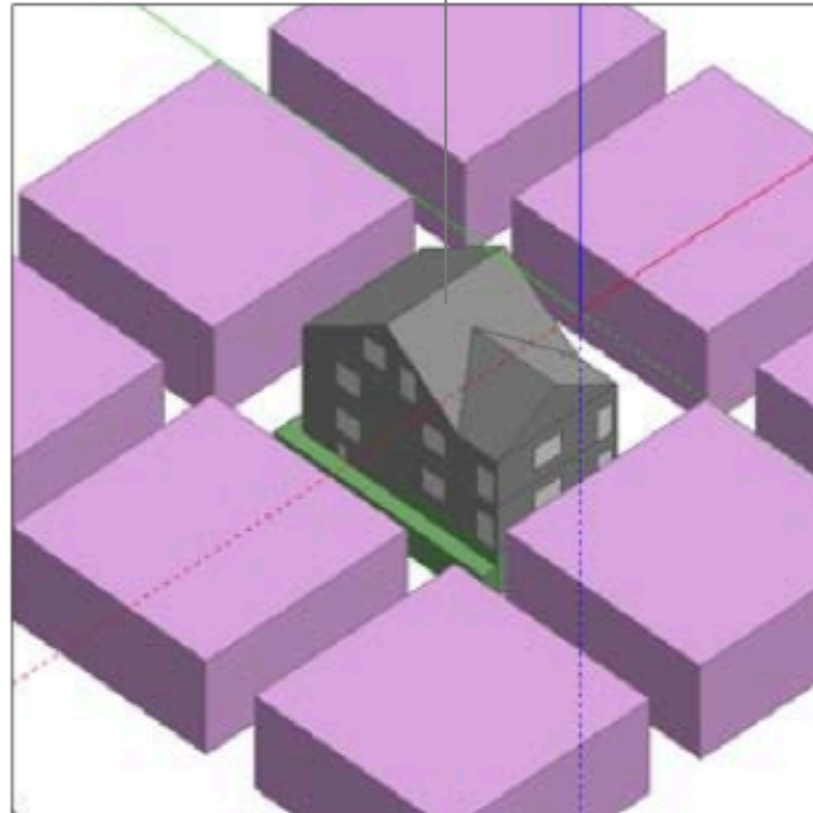
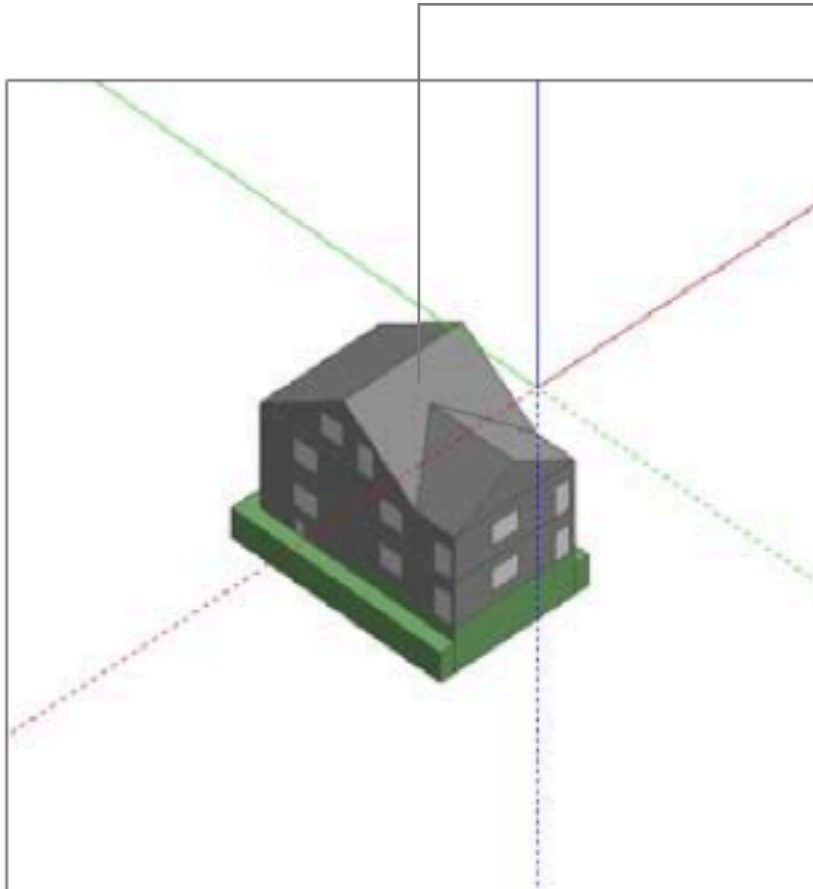
simulating
12 archetypes
with
5 different building volumes
5 different shading factors
5 different population inputs
are

180 simulations in BEM

simulating
11'000 buildings individually
are

11'000 simulations in BEM with
individual inputs

Same building archetype



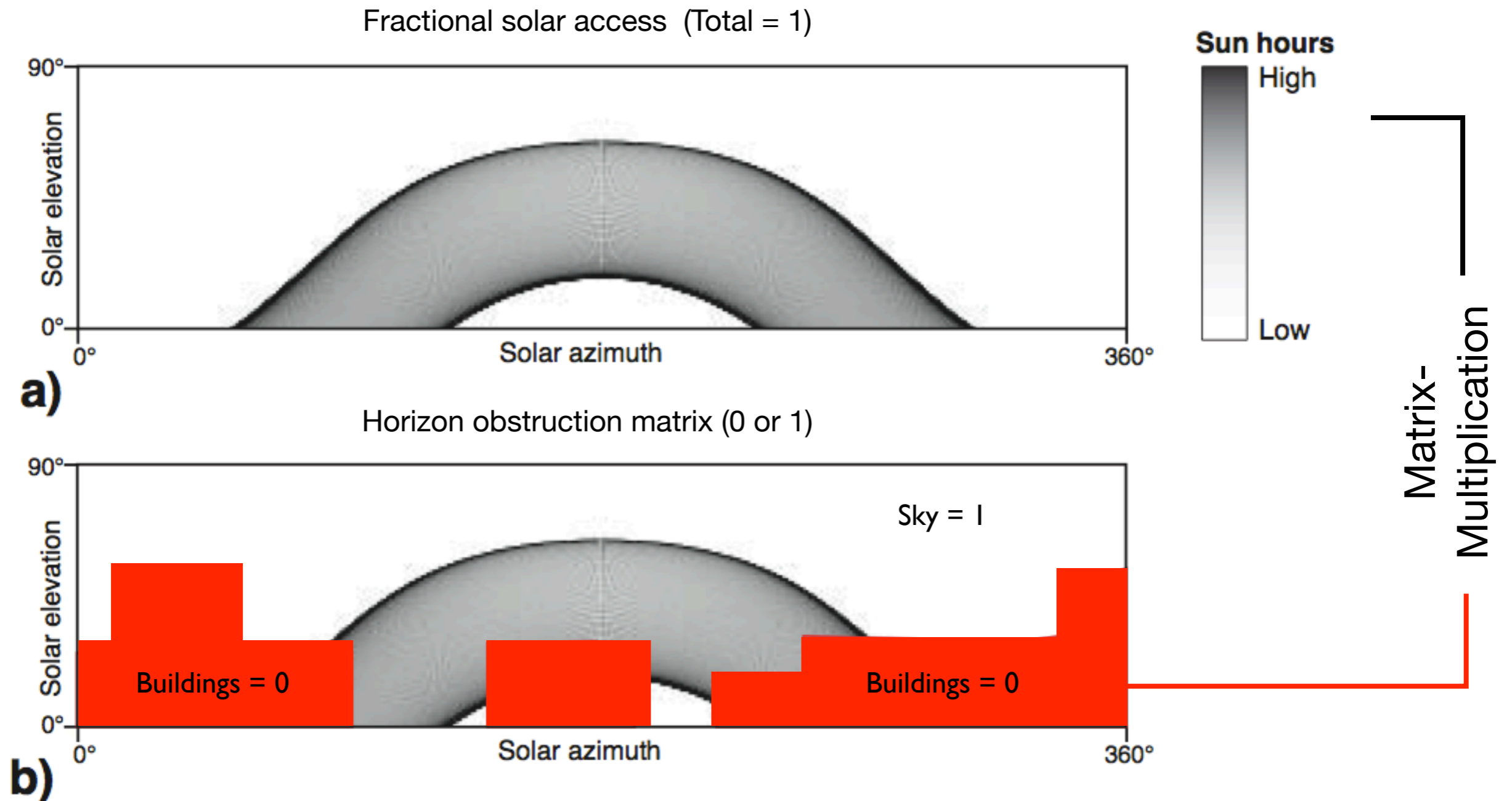
Reference BEM run
in unobstructed
terrain



4 BEM runs with
increasing
obstruction

Fractional shading

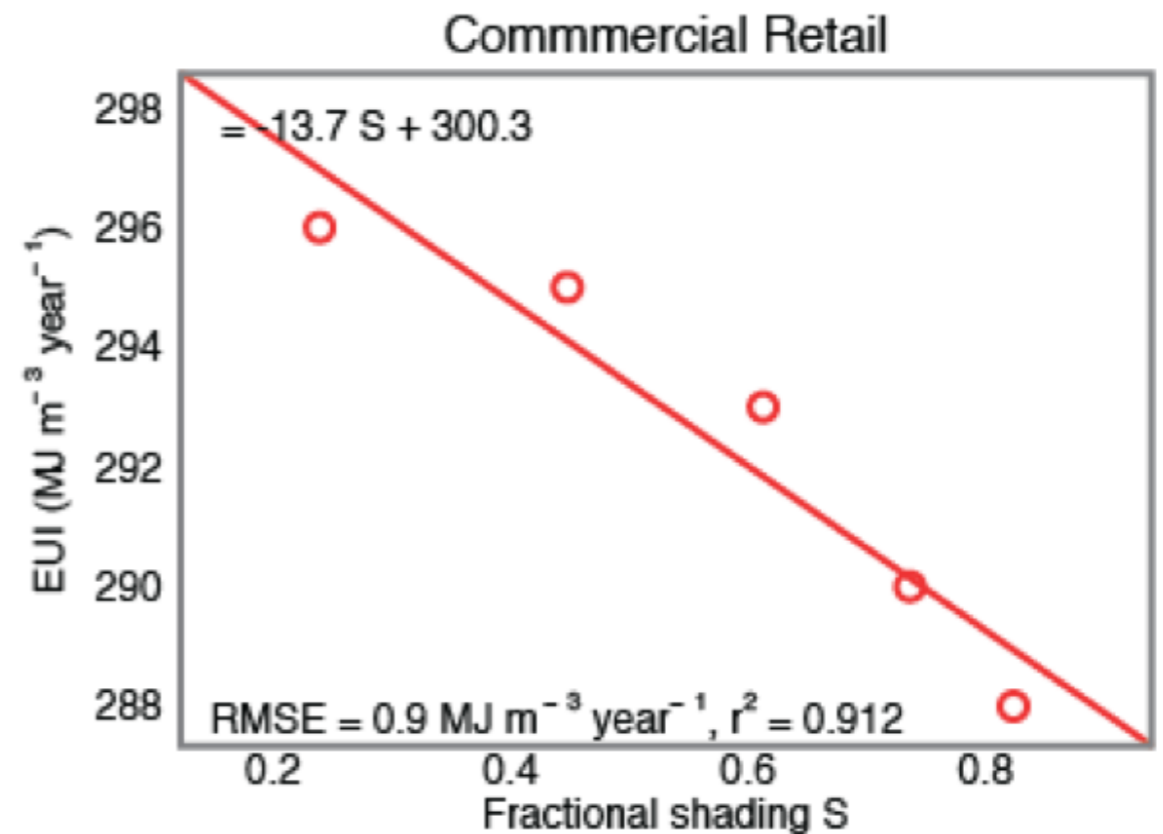
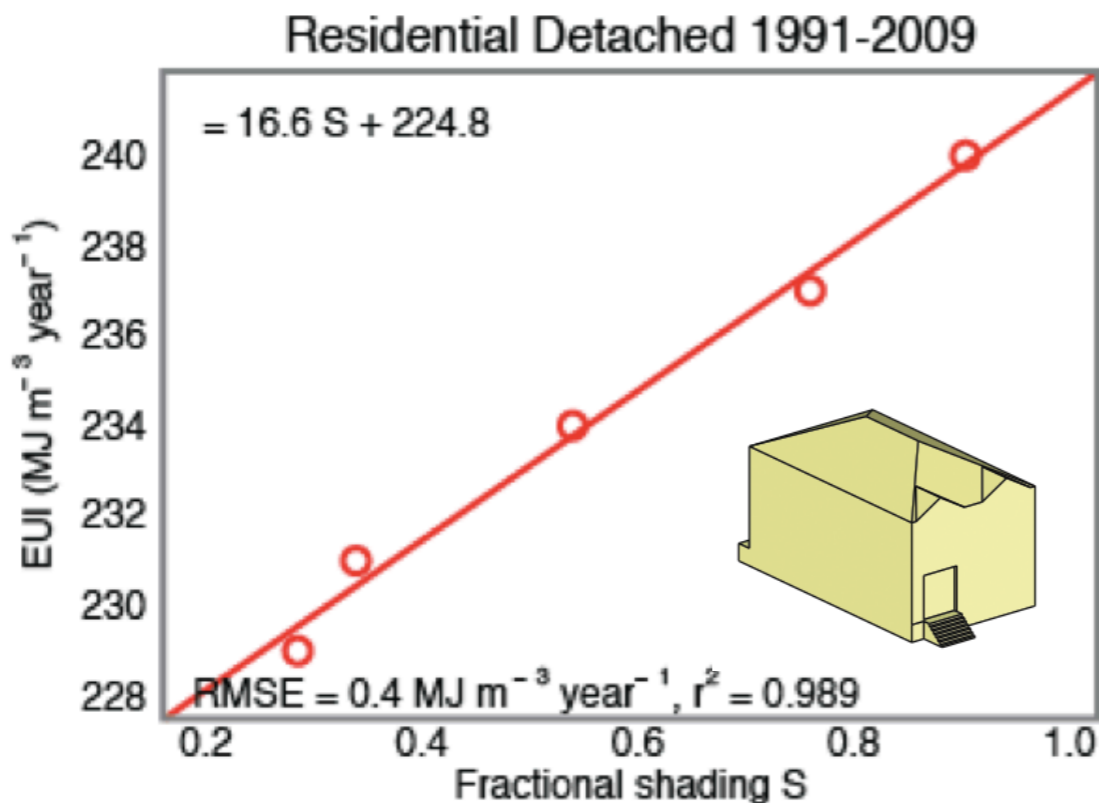
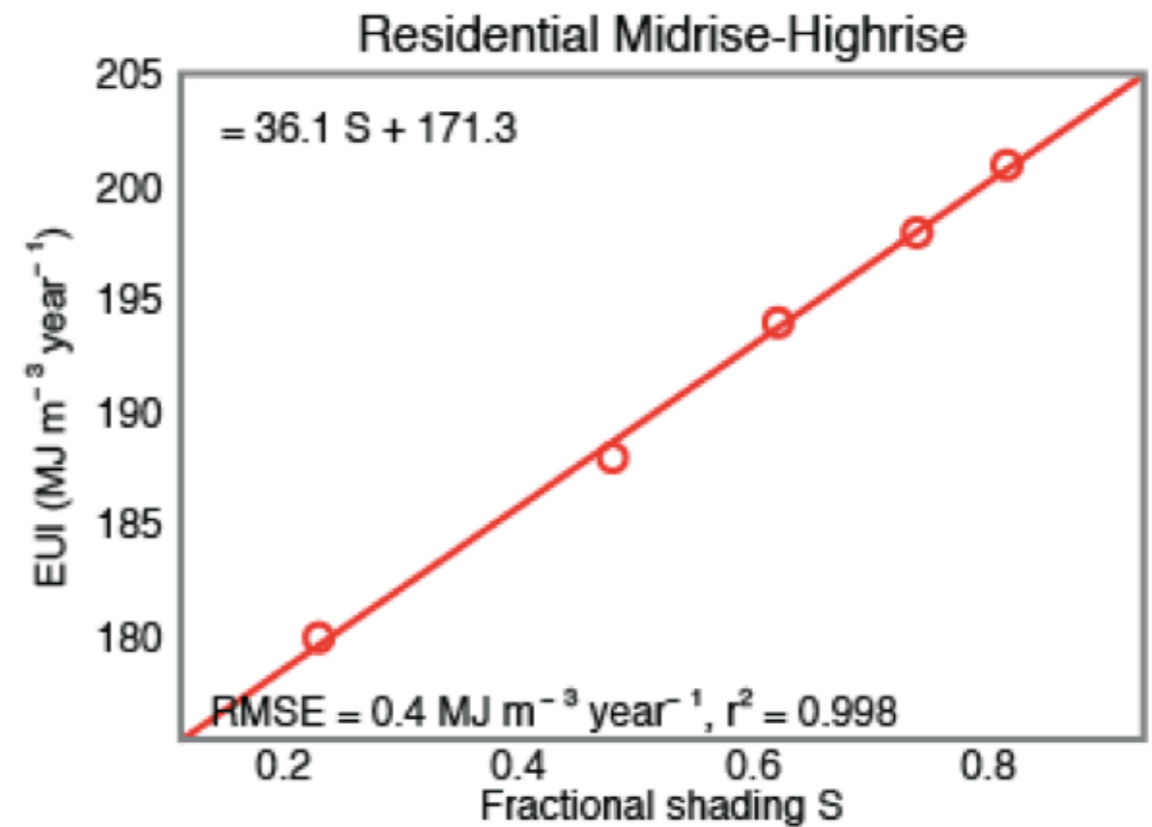
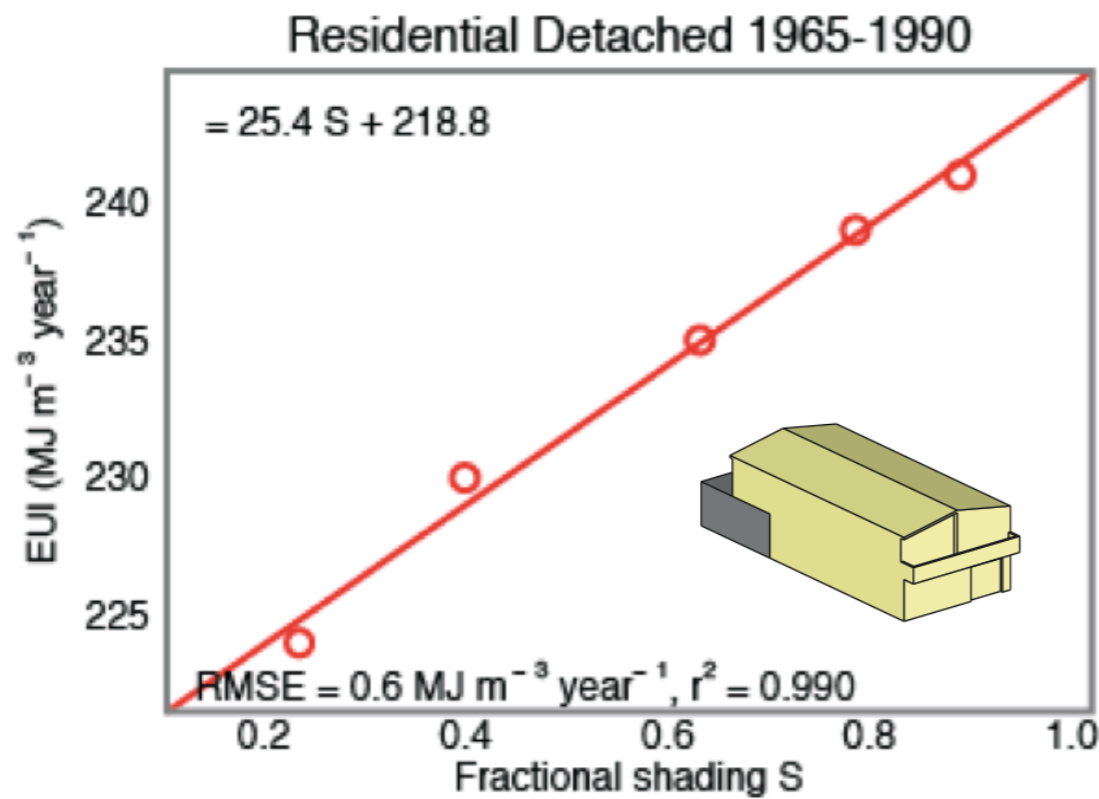
Rapid calculation of potential direct sunlight



Fractional shading calculated for for each building based on gridded LIDAR dataset

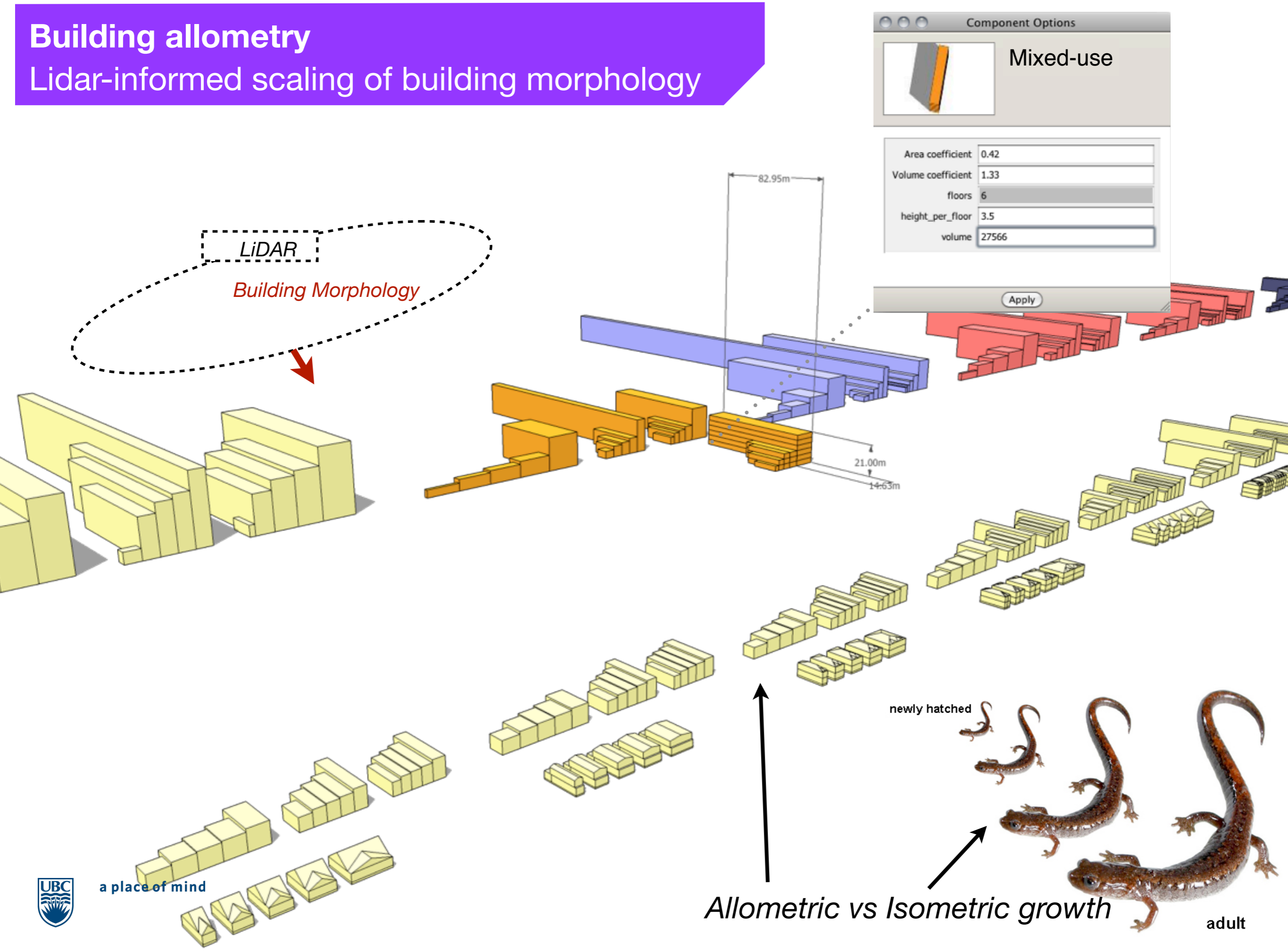
Fractional shading

Impact on Energy Use Intensity ($\text{MJ m}^{-3} \text{ year}^{-1}$)

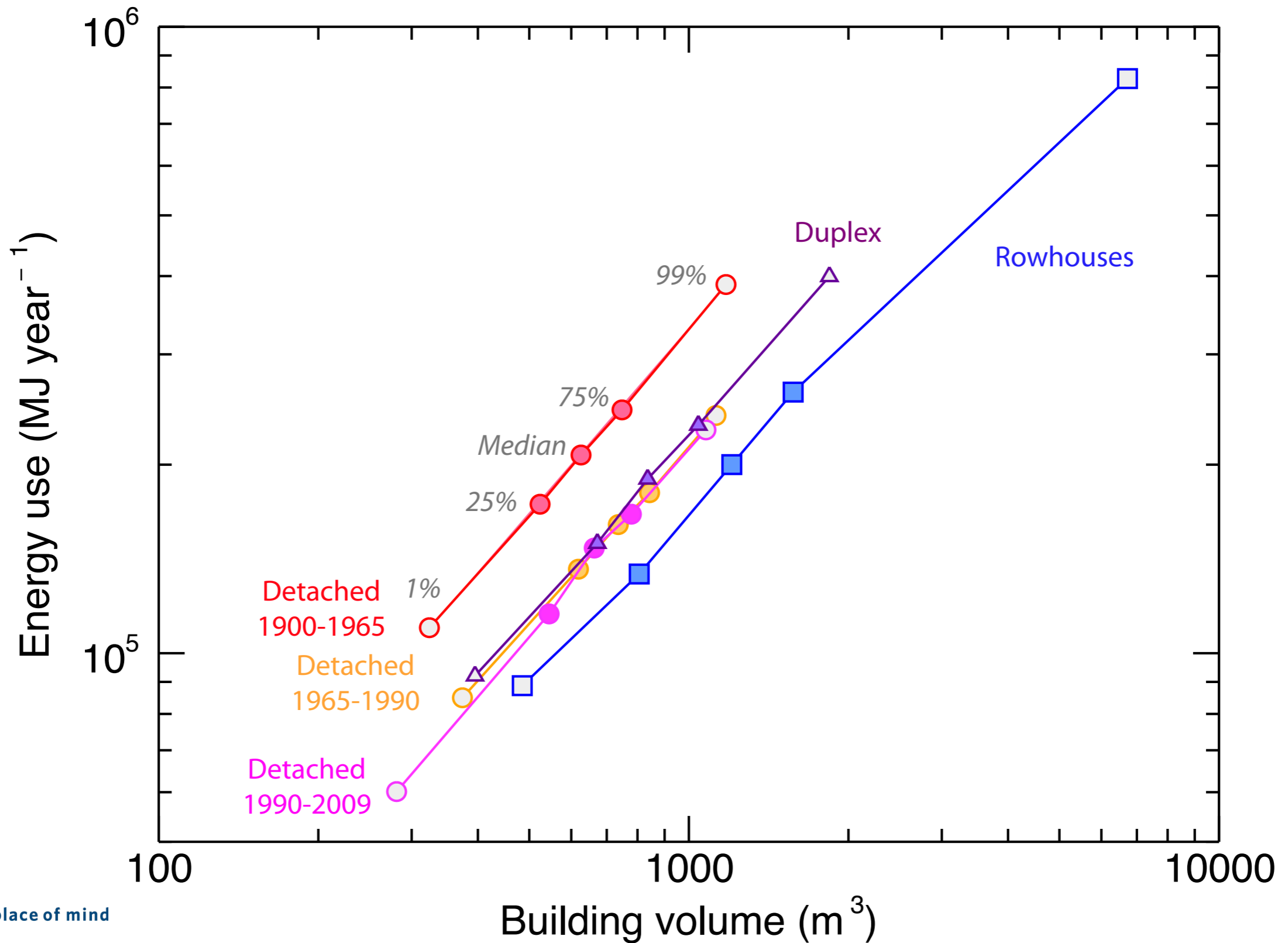


Building allometry

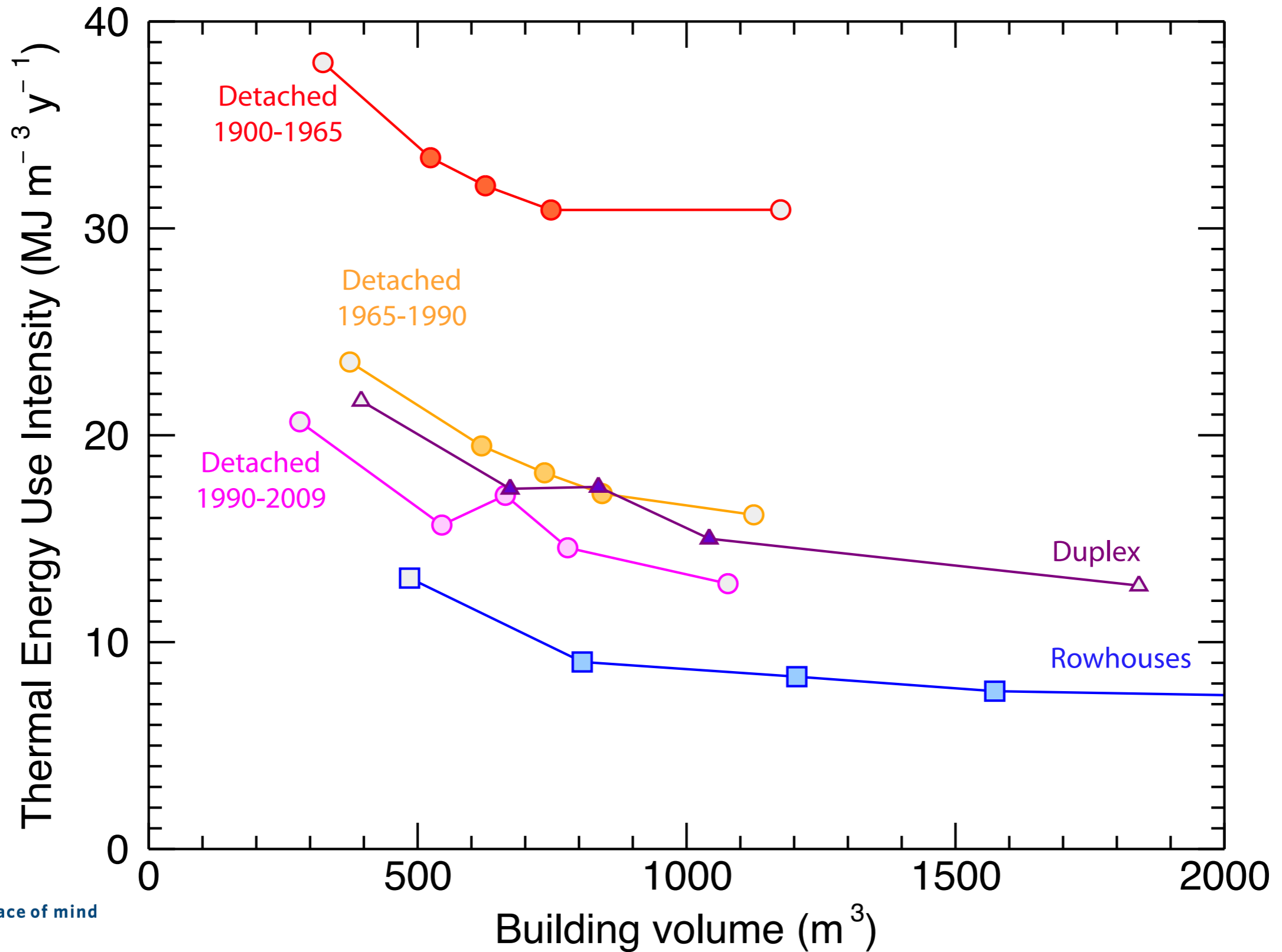
Lidar-informed scaling of building morphology



Total building energy use increases with volume

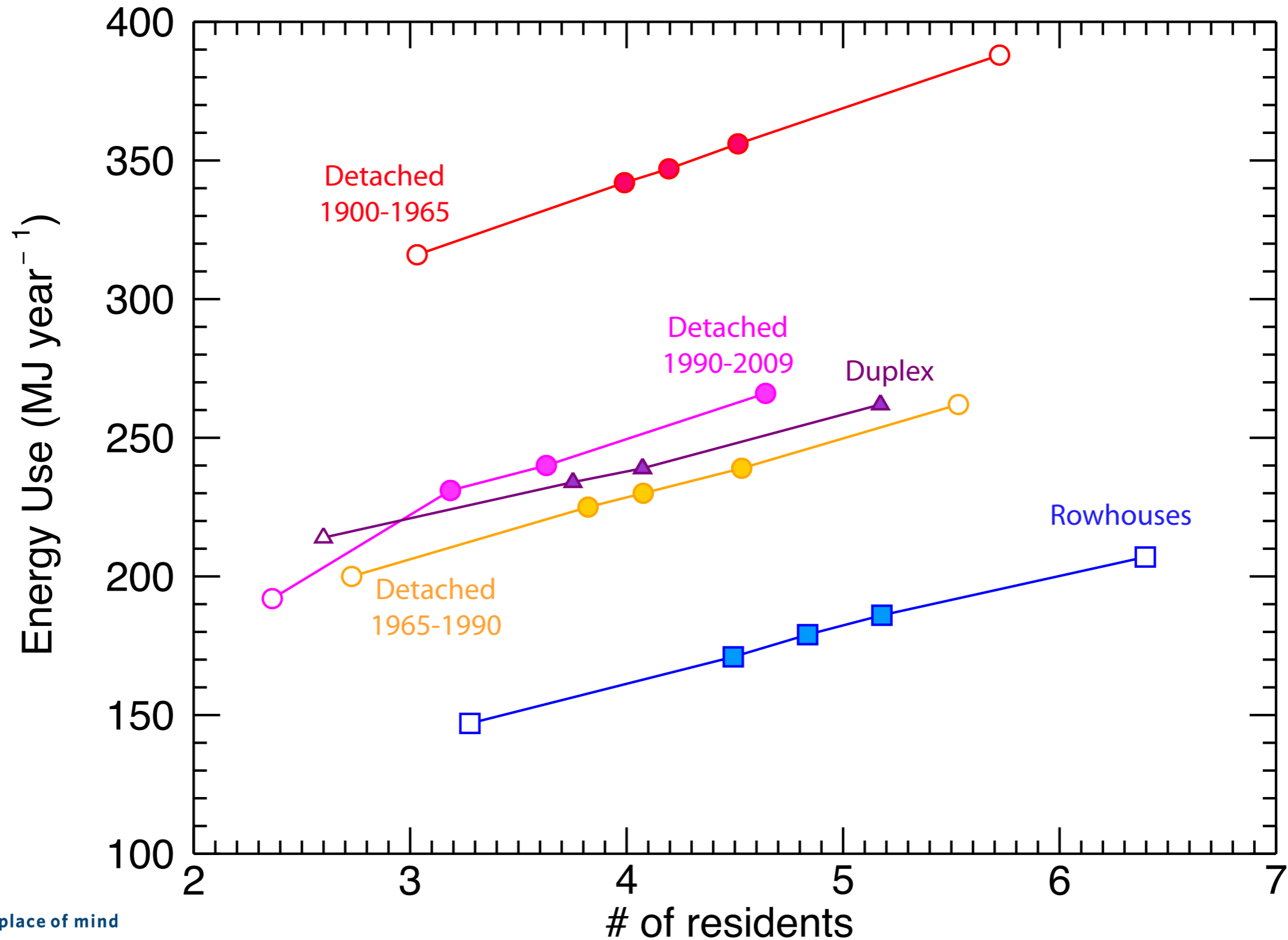


Energy intensity (m^{-3}) for space heating decreases with increasing volume



Population

driven primarily by hot water demand



$$EUI = \alpha_s S + \alpha_v V + \alpha_p (P - \bar{P}) + \beta_v$$

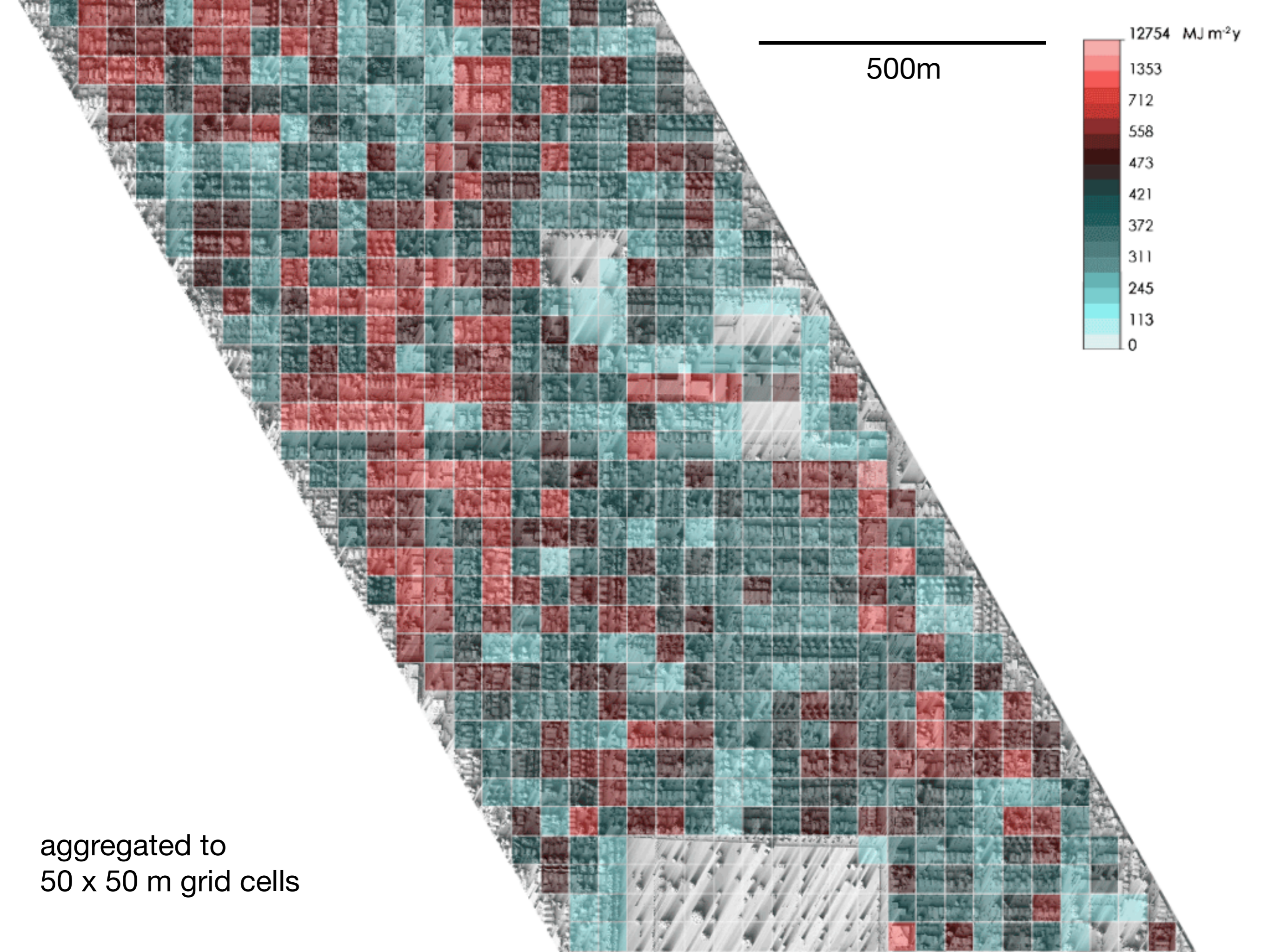
Urban Morphology Building Morphology Population Base-load

Sensitivity to shading
 ★ Fractional shading factor

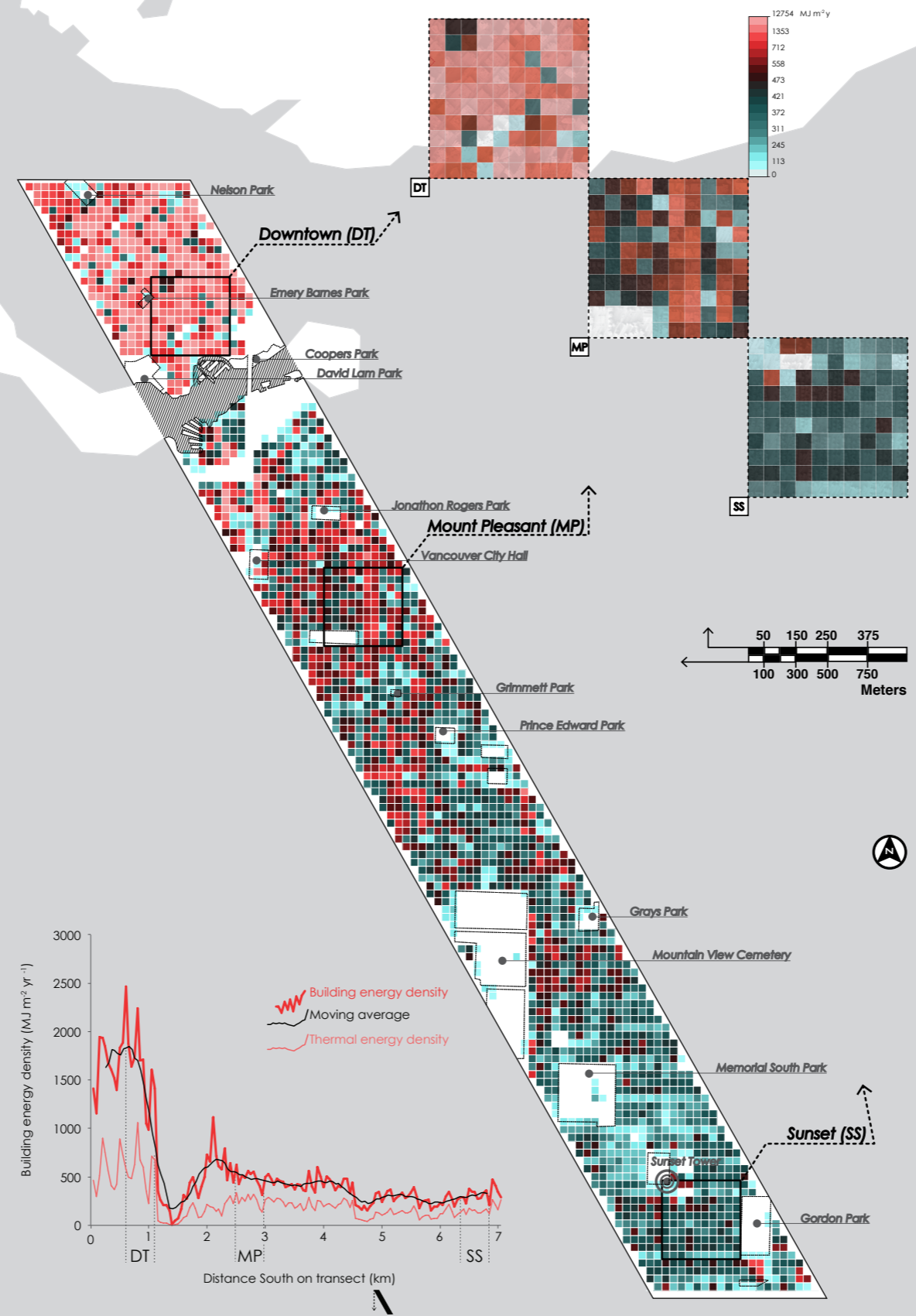
Sensitivity to volume
 ★ Building volume

Sensitivity to population
 ★ Population
 Average population of archetype

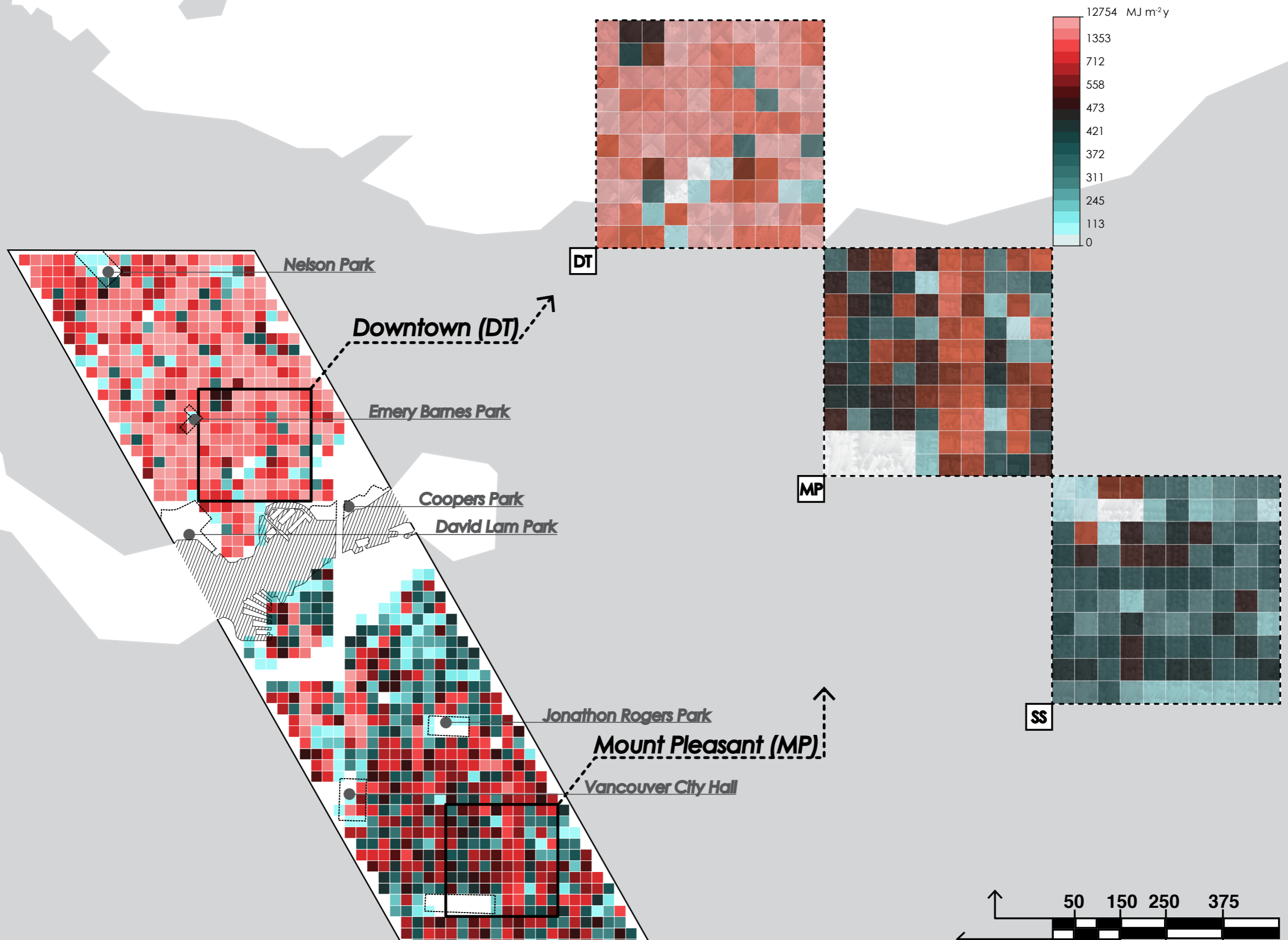


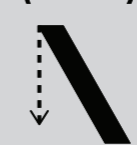
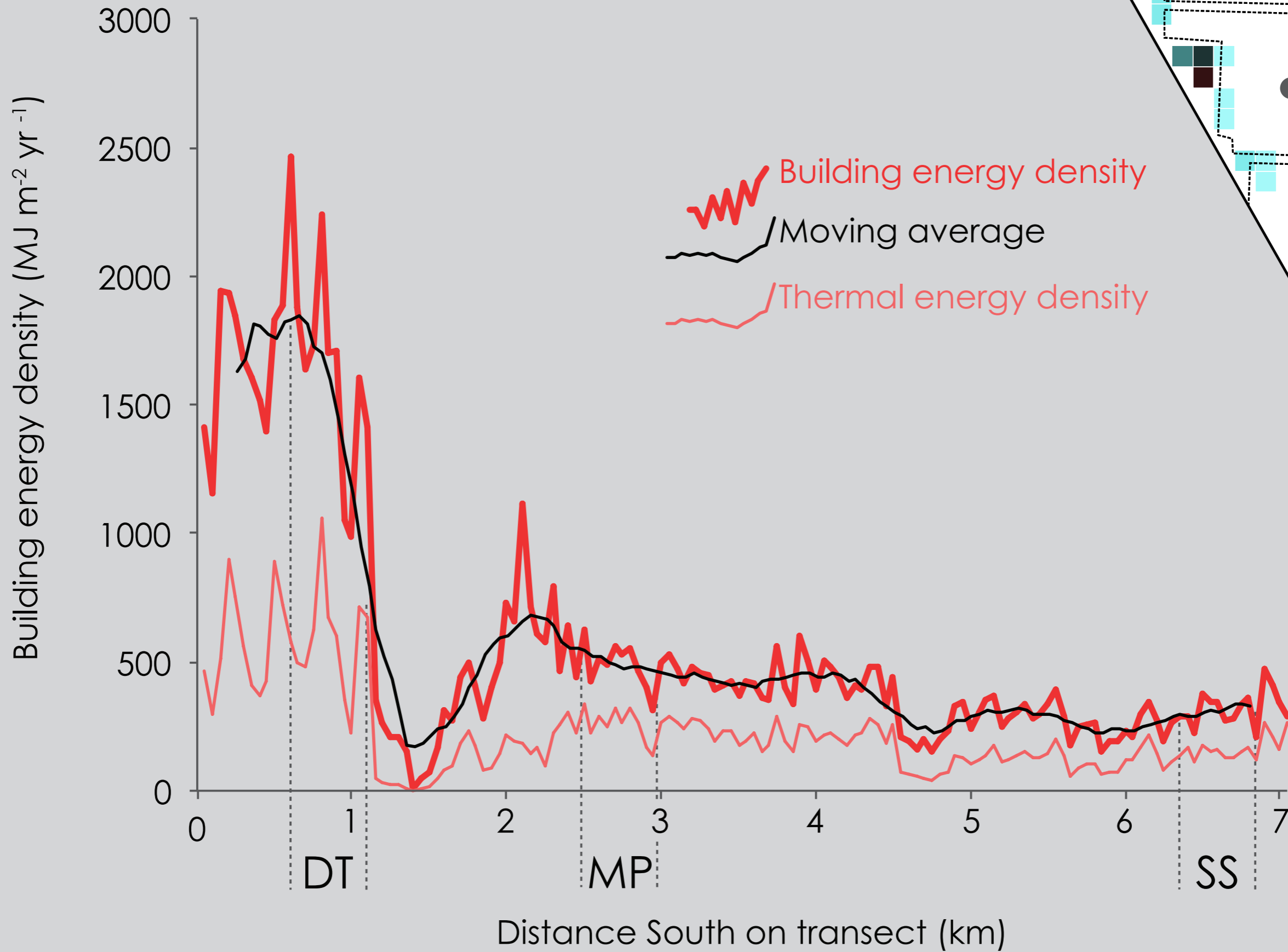


Building Energy Use

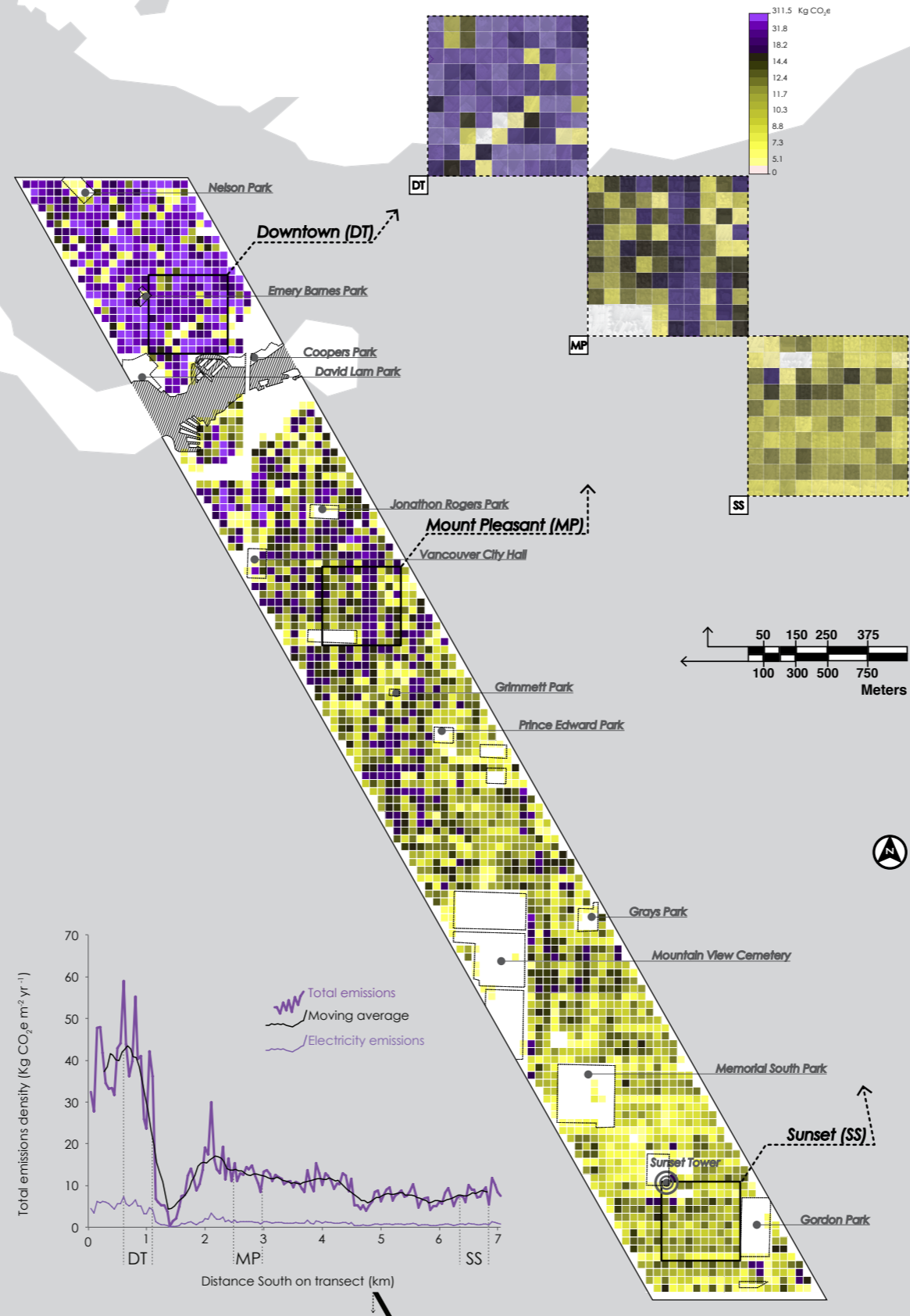


Building Energy Use





CO₂e Emissions



Conclusions and Summary

- Building energy demand has been **scaled using a multiple linear regression** model based on parameters derived from a **limited number of generic runs in a BEM**.
- Settings of the generic BEM runs were informed by **form data extracted from LIDAR**, and an **archetype database** that incorporates actual buildings and energy audits.
- Inputs to linear regression equation for each individual building were efficiently extracted from gridded LIDAR and land-use data.
- Approach enables data-products on **city or neighborhood-scale**, usable for urban planning, load forecasting, emission reduction and atmospheric models (emissions, anthropogenic heat flux).